

Increasing intensity of extreme global heatwaves: the crucial role of metrics

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Key Points:

- ERA5-Land is in good agreement with Berkeley-Earth and JRA-55 only over part of the Northern Hemisphere for daily maximum temperatures
- The most intense heatwaves of 1950-2021 change if considering intensity indices either based on cumulative or averaged values
- The most intense heatwaves of 1950-1985 have become up to ten times more usual and up to three times more intense during recent years

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Abstract

Many indices have been defined to estimate the intensity of a heatwave. However, these indices are often used indiscriminately, without sufficient consideration of their possible different results and of the challenges that this poses to a proper characterization and comparison of events. This study, by comparing four different indices applied to reanalyses data, shows that the choice of heatwave intensity metrics has important effects on the detection of the most intense events for the period 1950-2021, with indices based on cumulative values of a target variable that must be preferred over the ones relying on temporal averages. Under these considerations, one of the given indices is additionally selected for the study of heatwaves of the period 1950-2021, showing that heatwaves that were unlikely before 1986 have become up to ten times more usual and up to three times more intense during recent times.

Plain Language Summary

This work sets the basis for a more consistent and unified use of metrics for the assessment of heatwave intensity. Following the evidence of previous studies, here we confirm that the way heatwave intensity is calculated might lead to completely different outcomes, such as in the case of the most intense events occurring globally over the period 1950-2021: indices of heatwaves magnitude have to be based on cumulative values of the anomalies of a target variable rather than on temporal averages. Additionally, considering a metric based on cumulative values of standardized anomalies of daily maximum temperatures, the trends of very extreme heatwaves over the period 1950-2021 show that for all of the considered regions, what was rarely recorded in the period 1950-1985 has become up to ten times more likely and up to three times more intense over the years 1986-2021.

1 Introduction

Heatwaves are defined as extended periods of extreme warm temperature anomalies (Perkins & Alexander, 2013; Perkins-Kirkpatrick & Lewis, 2020). They are considered to be one of the most harmful natural hazards, with serious implications for human health (Kovats & Kristie, 2006; Fischer & Schär, 2010; Williams et al., 2012; Cusack et al., 2011; López-Bueno et al., 2021), infrastructure (Forzieri et al., 2018; Maggiotto et al., 2021; Stone Jr et al., 2021), the economy (García-León et al., 2021) and natural ecosystems (Breshears et al., 2021).

Heatwaves are normally assessed through measures of their intensity, frequency, duration and spatial extent (Perkins-Kirkpatrick & Lewis, 2020). There exists a plethora of different metrics for characterizing each of these features, often tailored to the specific needs of a given study, depending on the sector and area of the application. This heterogeneity in the use of metrics does not always allow for a comprehensive understanding of how heatwaves differ over time as well as by region. For this reason, many studies have called for a unified and consistent way of defining heatwaves (Russo & Sterl, 2011; Perkins & Alexander, 2013; Russo et al., 2015; Perkins-Kirkpatrick & Lewis, 2020).

One parameter of particular importance for the characterization of heatwaves and their impact is the heatwave magnitude or intensity. This heatwave characteristic is relevant since it is directly linked to the severity of heatwave impacts on natural ecosystems (Iwasaki & Noda, 2018). To assess the magnitude of a heatwave, a wide range of indices has been proposed that can be classified into two groups: 1. considering metrics based on temporal averages of a target variable (Cowan et al., 2014; Holbrook et al., 2022; Schaeffer & Roughan, 2017; Yu et al., 2020; Cueto et al., 2010; Perkins et al., 2012); 2. relying on cumulative values of the anomalies of a given variable calculated over the duration of an event (Russo & Sterl, 2011; Russo et al., 2014, 2015, 2016). The intensity

62 of a heatwave for a specific location is related to the duration of the event (Russo et al.,
 63 2014). While metrics based on cumulative values consider both the magnitude and the
 64 duration of heatwaves jointly, averaging does not allow for a direct comparison of heat-
 65 waves of different length. However, metrics for the characterization of heatwave inten-
 66 sities based on cumulative or averaged values are often used indiscriminately. There is
 67 therefore a need to better assess the effects of the two approaches, highlighting possi-
 68 ble differences that might alter any conclusion relevant for the detection, prediction and
 69 understanding of heatwaves.

70 In this study, the most intense heatwaves occurring over the period 1950-2021 are
 71 detected and characterized at a global scale by means of four different heatwave mag-
 72 nitude indices either based on cumulative or averaged values of temperature-based vari-
 73 ables. The main goal of this study is to identify possible inconsistencies between the two
 74 families of metrics and to set the basis for a more consistent and unified use of indices.
 75 All the presented analyses are conducted on the ERA5-Land reanalysis dataset (Muñoz-
 76 Sabater et al., 2021). Prior to the index calculation, ERA5-Land is evaluated against the
 77 Berkeley-Earth and JRA-55 datasets, to identify areas where it can be considered more
 78 reliable in terms of interannual variability of daily maximum temperatures. Then, the
 79 maximum heatwaves intensity and the year in which these events occur over the period
 80 from 1950 to 2021, according to ERA5-Land, are determined using the four considered
 81 indices, with the goal of highlighting different conclusions arising from the use of differ-
 82 ent approaches. Finally, in a last step, differences between the first and the second half
 83 of the considered study period in terms of occurrence and magnitude of very extreme
 84 heatwaves are investigated for one of the proposed metrics and specific regions.

85 2 Methods

86 2.1 Data

87 The analyses presented in this study are based on daily maximum temperature for
 88 the period 1950-2021 from the ERA5-land reanalysis dataset on a regular grid at a spa-
 89 tial resolution of 0.25° longitude \times 0.25° latitude.

90 ERA5-Land provides hourly information of surface variables at a spatial resolution
 91 of ~ 9 km. The data is derived from a single simulation with the ECMWF *Carbon Hydrology-
 92 Tiles scheme for Surface Exchanges over Land (CH-TESSSEL)* model, forced by mete-
 93 orological fields of the lowest atmospheric level of the ERA5 reanalysis (Hersbach et al.,
 94 2020), with an additional lapse-rate correction (Muñoz-Sabater et al., 2021). The model
 95 version employed for the production of ERA5-Land is very similar to the one used for
 96 ERA5, but with an updated parameterization of the soil thermal conductivity after Peters-
 97 Lidard et al. (1998), technical fixes improving the conservation of soil moisture balance
 98 and additional improvements for the calculation of potential evapotranspiration fluxes.
 99 These improvements do not lead to remarkable differences between ERA5-Land and ERA5,
 100 given the fact that they still share common and similar parameterizations of land pro-
 101 cesses (Muñoz-Sabater et al., 2021). The main added value of ERA5-Land over ERA5
 102 is attributable, according to Muñoz-Sabater et al. (2021), to the non-linear dynamical
 103 downscaling with corrected thermodynamic input, allowing for example to better dis-
 104 criminate between land and sea points over coastal areas.

105 The reliability of ERA5-Land in terms of daily maximum temperatures is assessed
 106 here against two additional datasets: the gridded Berkeley-Earth observational dataset
 107 (Rohde et al., 2013) (BE hereafter) and the Japanese reanalysis dataset JRA-55 (Kobayashi
 108 et al., 2015). These datasets are chosen as they have a temporal coverage similar to ERA5-
 109 Land, with BE starting in 1950 and JRA-55 in 1958. Daily gridded values of maximum
 110 temperatures are available on a regular grid with a spatial resolution of 1° longitude \times
 111 1° latitude for BE and of 1.25° longitude \times 1.25° latitude for JRA-55. For the comparison

112 against these two other datasets, ERA5-Land is first upscaled onto the respective coarser-
113 resolution grids from each dataset, through conservative remapping.

114 All the employed datasets cover the entire globe. However, in this study only the
115 data between -80° to 80° N are considered. Additionally, it is important to note that for
116 the comparison of ERA5-Land against JRA-55, daily maxima are obtained from 6-hourly
117 instead of from 1-hourly data, corresponding to the JRA-55 temporal resolution.

118 2.2 Heatwave Definition

119 Heatwave events are defined here as at least 3 consecutive days with temperatures
120 exceeding a given threshold. Similar to the definition used in Russo et al. (2015) and Perkins-
121 Kirkpatrick and Lewis (2020), for a specific day d , here the threshold $Tr90_d$ is defined
122 as the 90th percentile of daily maximum temperatures, in a sliding window of 30 days
123 around the considered day of a year, over a 30-year reference period.

124 We select the period from 1961 to 1990 as our reference period, as this period is
125 often considered as a reference for long-term climate change assessments (Tavakol et al.,
126 2020).

127 2.3 Heatwave Magnitude Indices

128 With the goal of identifying possible differences arising from the application of met-
129 rics using cumulative or averaged values of temperature-based variables, four different
130 ways of assessing heatwave intensity are considered here, two for each of the two given
131 classes. The first one is based on the magnitude assessment of single heatwave events over
132 a season, while the other three jointly consider all the days characterized by a heatwave
133 during an entire season. Below, the four indices are described in detail.

134 2.3.1 *HWMI_d*

135 The *HWMI_d* of Russo et al. (2015) is calculated as the sum, for a single point, of
136 the daily magnitude index (M_d) over each of the days composing a heatwave event. For
137 computing M_d , first anomalies of daily maximum temperatures for a given day d are com-
138 puted with respect to the 25th percentile of yearly maxima over the reference period. Then,
139 the anomalies are standardized by the interquartile range (IQR) of the yearly maxima
140 of daily maximum temperatures over the reference period, allowing for a comparison of
141 different points in space characterized by different interannual variability:

$$M_d(T_d) = \begin{cases} \frac{T_d - T_{30y25p}}{T_{30y75p} - T_{30y25p}} & \text{if } T_d > T_{30y25p} \\ 0 & \text{if } T_d \leq T_{30y25p} \end{cases} \quad (1)$$

142 where T_d is the daily maximum temperature on day d of a heatwave, T_{30y25p} and
143 T_{30y75p} are, respectively, the 25th and 75th percentile values of the time-series composed
144 of 30-year yearly maxima of daily temperatures for the reference period 1961–1990.

145 The methodology introduced by Russo et al. (2015) was designed for characteriz-
146 ing heatwaves over Europe, where the annual maxima of daily temperatures generally
147 occur in boreal summer. Here, considering almost the entire globe, it is important to ac-
148 knowledge that over other areas yearly maxima might take place at different times of the
149 year. Therefore, similarly to Russo et al. (2016), the presented analyses are conducted
150 separately for each season of the year (e.g., June July August (JJA)). In order to include
151 heatwaves that start in a season and finish in another, periods of five months are selected
152 around each 3-month season for the definition of heatwaves, with an additional month
153 at the beginning and at the end of their classical definition (e.g. May to September (MJ-
154 JAS) for boreal summer). Then, to avoid counting single events twice, heatwaves are as-

155 signed to a specific season depending on the largest number of days they have in the three
 156 central months of each season.

157 **2.3.2 Cumulative Heat**

158 The cumulative heat defined in Perkins-Kirkpatrick and Lewis (2020) is given by
 159 the sum of the anomalies with respect to the threshold in daily maximum temperatures
 160 of section 2.2, over all days characterized by a heatwave in each of the seasons s of the
 161 considered study period:

$$HEATcum_{y_s} = \sum_{d=1}^{n_{y_s}} T_d - Tr90_d \quad (2)$$

162 where y indicates the given year, d the heatwave day of a season, n the total num-
 163 ber of heatwave days in that season and $Tr90_d$ the 90th percentile threshold for a given
 164 day, as defined above.

165 **2.3.3 AVI**

166 The heatwave average intensity (AVI) of Perkins-Kirkpatrick and Lewis (2020) is
 167 the average temperature calculated over all the heatwave days of a season, for each year
 168 of the considered period:

$$AVI_{y_s} = \frac{\sum_{d=1}^{n_{y_s}} T_d}{n_{y_s}} \quad (3)$$

169 where, similarly to equation 2, y and s are, respectively, the considered year and
 170 season, d the given day of a heatwave, and n_{y_s} the total number of heatwave days in that
 171 season.

172 **2.3.4 AVA**

173 The heatwave Average Anomalies (here referred to as AVA) index is derived from
 174 the study of Perkins-Kirkpatrick and Lewis (2020) and represents the average of the tem-
 175 perature anomalies with respect to the corresponding threshold, calculated over all the
 176 heatwave days of a given season:

$$AVA_{y_s} = \frac{\sum_{d=1}^{n_{y_s}} T_d - Tr90_d}{n_{y_s}} \quad (4)$$

177 where, again, y and s are, respectively, the considered year and season, d is the given
 178 heatwave day and n_{y_s} the total number of heatwave days in that season.

179 **3 Results**

180 Prior to the comparison of the considered indices applied to daily temperature max-
 181 ima derived from ERA5-Land, an evaluation of ERA5-Land against BE and JRA-55 is
 182 conducted in terms of the interannual variability of seasonal maxima of the target vari-
 183 able for each grid point of the domain. This allows us to better understand where the
 184 data can be considered more reliable for the estimation of the most intense heatwaves
 185 of a given period.

186 Fig. 1 shows global maps of the Spearman rank correlation calculated over each
 187 grid point of the domain at a 1°longitude × 1°latitude spatial resolution, between the

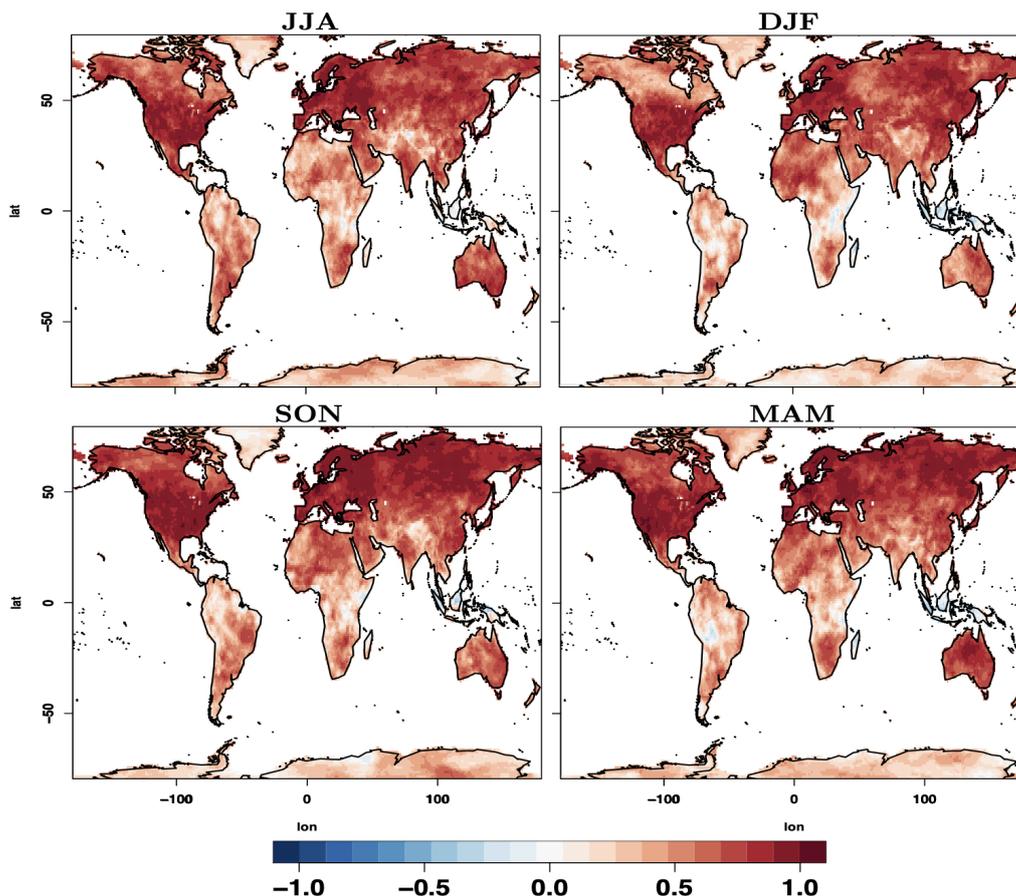


Figure 1. Spearman correlation calculated for the amplitude of the seasonal maxima of daily maximum temperatures between ERA5-Land and Berkeley Earth, over the period 1950-2021, at $1 \times 1^\circ$ resolution. The correlation is calculated for each season separately (from top to bottom: JJA, DJF, SON and MAM).

188 time-series of seasonal maxima of daily maximum temperatures derived from ERA5-Land
 189 and BE, for all four seasons of a year. In all seasons, the two datasets show a good agree-
 190 ment in terms of the considered variable between 30°N and 65°N , while the agreement
 191 between the datasets in the continental Southern Hemisphere is lower. For the tropical
 192 regions of South America and Africa, even negative correlations are evident in some cases
 193 between the two datasets. Regions characterized by complex topography, such as Green-
 194 land, Antarctica and the Tibetan plateau, also show very low correlation between ERA5-
 195 Land and BE, with values generally lower than $+0.2$ for all seasons. Northern North Amer-
 196 ica exhibits a remarkably low correlation (below $+0.5$) in DJF. Also, Western Australia
 197 shows a lower correlation in DJF and SON than in the other seasons. The same anal-
 198 yses conducted between ERA-Land and JRA-55 produce similar results, with Africa, South
 199 America and Antarctica presenting correlations lower than 0.5 for almost all grid points,
 200 and in all seasons (see supplements, Fig. S1). This agrees with findings comparing global
 201 daily maximum temperatures from ERA5 against JRA-55 (Thompson et al., 2022).

202 In a next step, daily maximum temperatures from the ERA5-Land dataset at a spa-
 203 tial resolution of $0.25^\circ\text{longitude} \times 0.25^\circ\text{latitude}$ are used to determine the most intense
 204 heatwaves of the period 1950-2021 and the year in which they occur, according to the

205 four indices defined in section 2.3. The goal is to investigate whether and how the de-
 206 tection of the most extreme events changes when considering different metrics. Fig. 2
 207 shows the maximum values of the four given indices, for each season separately, over the
 208 period 1950-2021. The points for which the correlation between seasonal maxima of daily
 209 maximum temperatures calculated between ERA5-Land and BE is lower than +0.5 are
 210 shaded in gray. Fig. 2 illustrates how the AVI has a completely different pattern of the
 211 maxima with respect to the other indices. This is due to the fact that the AVI consid-
 212 ers absolute temperatures: for this index it is not possible to properly compare extremes
 213 over regions characterized by different seasonal cycles (also relevant for heatwave pre-
 214 diction (De Perez et al., 2018)). Considering the indices based on temperature anoma-
 215 lies, the AVA, relying on temporal averages, has a different pattern of the maxima with
 216 respect to the two other indices (i.e., HWMId and HEATcum). In particular, the HWMId
 217 and HEATcum show maximum values over corresponding areas, in all seasons, except
 218 for SON when HEATcum shows much larger values over the high northern latitudes. In
 219 general, more pronounced maxima are evident over the higher latitudes of the North-
 220 ern Hemisphere for HEATcum than for HWMId. This is true also when considering the
 221 HWMId calculated over all the heatwave days of a season, as for HEATcum (see sup-
 222 plements, Fig. S2), confirming that these differences are due to the consideration of stan-
 223 dardized values of the anomalies for the HWMId. In JJA, HWMId and HEATcum both
 224 have some of their highest values over Western Russia, which can be associated with the
 225 extreme summer heatwave of 2010 (Russo et al., 2014, 2015). On the other hand, in JJA
 226 values of AVA over this region are not very pronounced and other areas show consider-
 227 ably larger values in terms of the defined metric. This behavior is noticeable also when
 228 considering the extreme values of HWMId and HEATcum for Central Africa and South
 229 America, in JJA, SON and MAM (where ERA5-Land exhibits a strong disagreement with
 230 respect to the other datasets): these anomalous events almost disappear in the case of
 231 AVA.

232 Another interesting way to look at possible differences arising from the application
 233 of the different metrics is by considering the years when the corresponding event with
 234 maximum magnitude occurs over the period 1950-2021. A very important result evident
 235 from Fig. 3 is that while the maps of the year when the maxima in the given metrics oc-
 236 cur are pretty similar in the case of HWMId and HEATcum, the AVI and AVA show a
 237 very different spatial distribution. While for the first two indices (shown in the first two
 238 columns), for more than 70% of the land areas the most intense heatwave event occurs
 239 during the last 36 years of the considered period, in the other two more than 50% of the
 240 most extreme events take place before the year 1986. In particular, the differences be-
 241 tween AVA and HEATcum are considerably larger than the differences between HWMId
 242 and HEATcum in all cases, even though HWMId is not only based on standardized anoma-
 243 lies, but it also considers cumulative values over single events instead of an entire sea-
 244 son. Hence, the conclusions on the most intense events and the years in which they oc-
 245 cur that can be drawn from the two groups of indices are substantially different. This
 246 demonstrates that metrics assessing heatwave intensity based either on temporal means
 247 or cumulative values cannot be used indiscriminately.

248 Finally, the trends in the intensity of single heatwave events are investigated over
 249 the entire period 1950-2021. The HWMId allows us to calculate heatwave intensity for
 250 single events, while at the same time providing a standardized measure useful for the com-
 251 parison across time and space. Hence, for the next analysis, the HWMId of single heat-
 252 wave events over the period 1950-2021 is calculated for selected subregions of the North-
 253 ern Hemisphere, namely Central North America (CNA), Europe (EUR) and Northern
 254 Asia (NAS, see supplements, Fig. S3), for which the ERA5-Land shows a better agree-
 255 ment with other datasets in terms of the interannual variability of daily maximum tem-
 256 peratures. Over each of these regions, for each season, the changes in the number of very
 257 extreme events and their intensities between two periods of 36 years (hereafter referred
 258 to as Early Period (EP) and Late Period (LP), respectively), the first starting in 1950

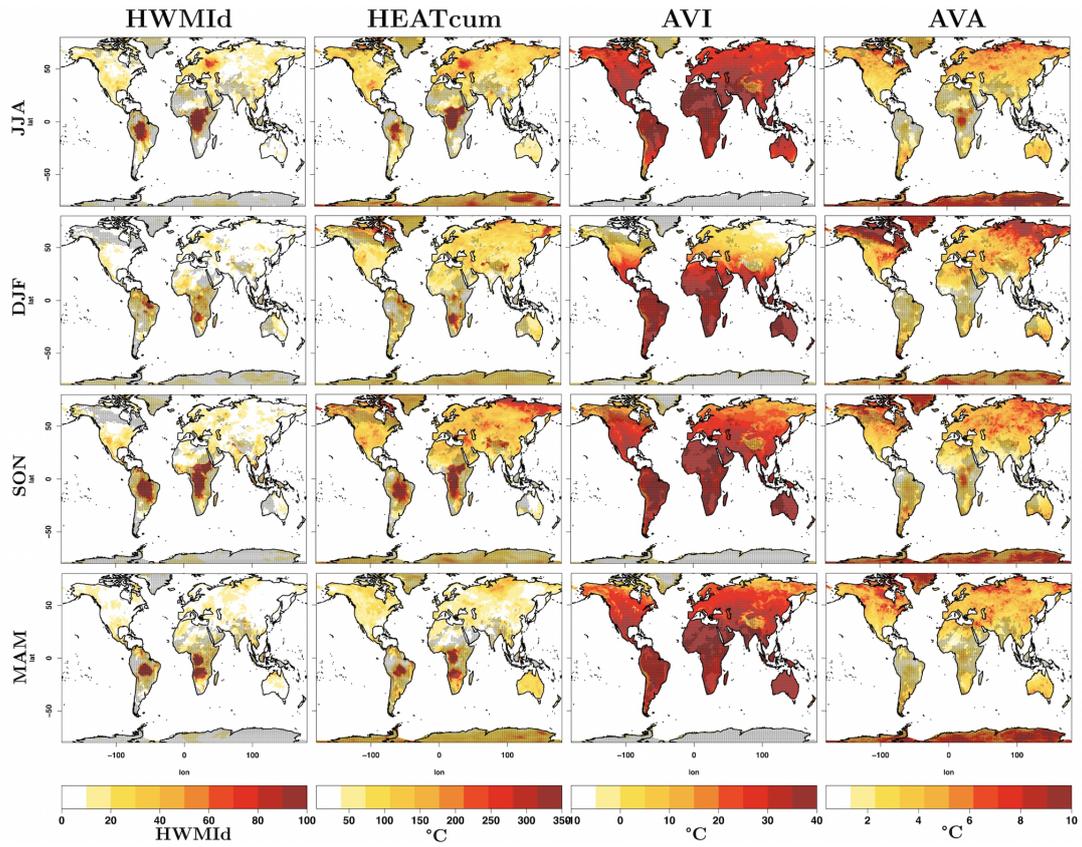


Figure 2. From top to bottom, maximum values of JJA, DJF, SON and MAM for (from left to right) HWMId, HEATcum, AVI and AVA, applied to ERA5-Land daily maximum temperatures over the period 1950-2021.

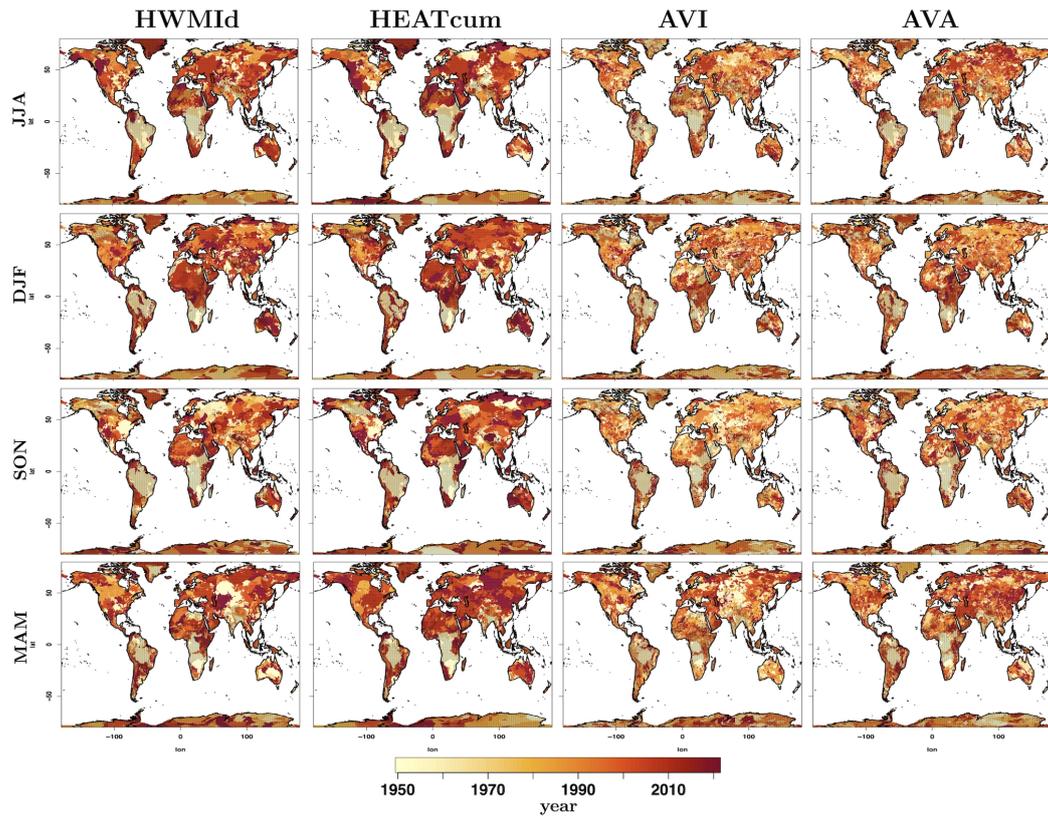


Figure 3. Year when the maximum values of, from left to right, HWMId, HEATcum, AVI and AVA occur over the period 1950-2021, according to ERA5-Land daily maximum temperatures. From top to bottom, the seasonal values derived for JJA, DJF, SON and MAM are shown.

and the second in 1986 (except DJF where periods of 35 years are considered, starting in 1952 and 1987, respectively) are investigated. A heatwave event is considered as very extreme when its corresponding HWMId value is larger than the 99.9th percentile of the values calculated for all the points of the considered subdomain over the period 1950-1985.

Fig. 4 shows the number of very extreme heatwaves for different values of HWMId, for all seasons and considered regions. In general, a higher number and more intense extreme heatwaves are evident over the period 1986-2021 compared to 1950-1985, in all seasons and regions. In CNA, the maximum intensities are higher during LP than in EP, for all seasons, up to almost two times in JJA. The number of very extreme events also strongly increases in this region during LP compared to EP, up to almost four times more in MAM. In CNA, during LP, only a small number of events exceeds the maximum intensity of heatwaves during EP, for a maximum number of 118 times in JJA. In NAS, the maximum intensities are also more pronounced during the most recent of the two periods, for all seasons, with values up to more than three times higher in JJA. For NAS the number of very extreme events increases by more than a factor of two over LP, in all seasons, with an exceptional increase by a factor greater than five in DJF and JJA. For the same region, during LP, a large amount of events exceeds the highest intensity of EP, for a maximum of 262 times in DJF and 599 in JJA. The largest changes between the two periods in both the number of events and their maximum magnitudes are evident for Europe, in particular in JJA, DJF and MAM. Here an event considered very extreme during EP occurs at least four times more often during the most recent period in JJA and MAM, and more than ten times more often in DJF. The maximum HWMId value registered over the two periods for EUR is almost the same in SON, approximately twice in DJF and up to three times more in MAM and JJA during recent times. Additionally, for EUR, in an exceptionally high number of cases the maximum intensity of the period 1950-1985 is exceeded during the most recent period, up to 761 times in DJF and almost 2000 times in JJA.

4 Conclusions

Several studies have called for a more unified definition and assessment of heatwave characteristics. Nonetheless, a plethora of different approaches is still employed for the study of heatwaves. In particular, concerning heatwave intensity, metrics based on cumulative or averaged values of a target variable are often used indiscriminately.

The results presented in this study show that the selection of metrics for the assessment of heatwave intensities needs extreme caution: the year and spatial distribution of the most intense events over the period 1950-2021, as calculated from daily maximum temperatures from the ERA5-Land reanalysis, change remarkably when considering four different indices belonging to two families of metrics, one based on temporal averages and the other on cumulative values. The use of metrics based on cumulative values should be preferred over the ones relying on temporal averages since, as already suggested by Russo et al. (2014), assessing intensity through averaged values does not allow for an unequivocal comparison of the magnitude of events with differing length. One simple example that could help in clarifying this point further is by considering two different heatwaves, the first one, HW1, lasting three days and the second, HW2, lasting four days. Supposing that HW1 has a value of the anomalies for each of the three heatwave days of +3°C, and HW2 has a value of the anomalies of +3°C for three heatwave days and of 2°C for the fourth one, when considering the average value of the anomalies the event HW1 will misleadingly be considered more intense than HW2. An additional important consideration on the reason to prefer cumulative values over averaged ones in the computation of heatwaves intensity is that, from an impact point of view, it is important to assess accumulated excess heat experienced over a given period of time (Perkins-Kirkpatrick & Lewis, 2020).

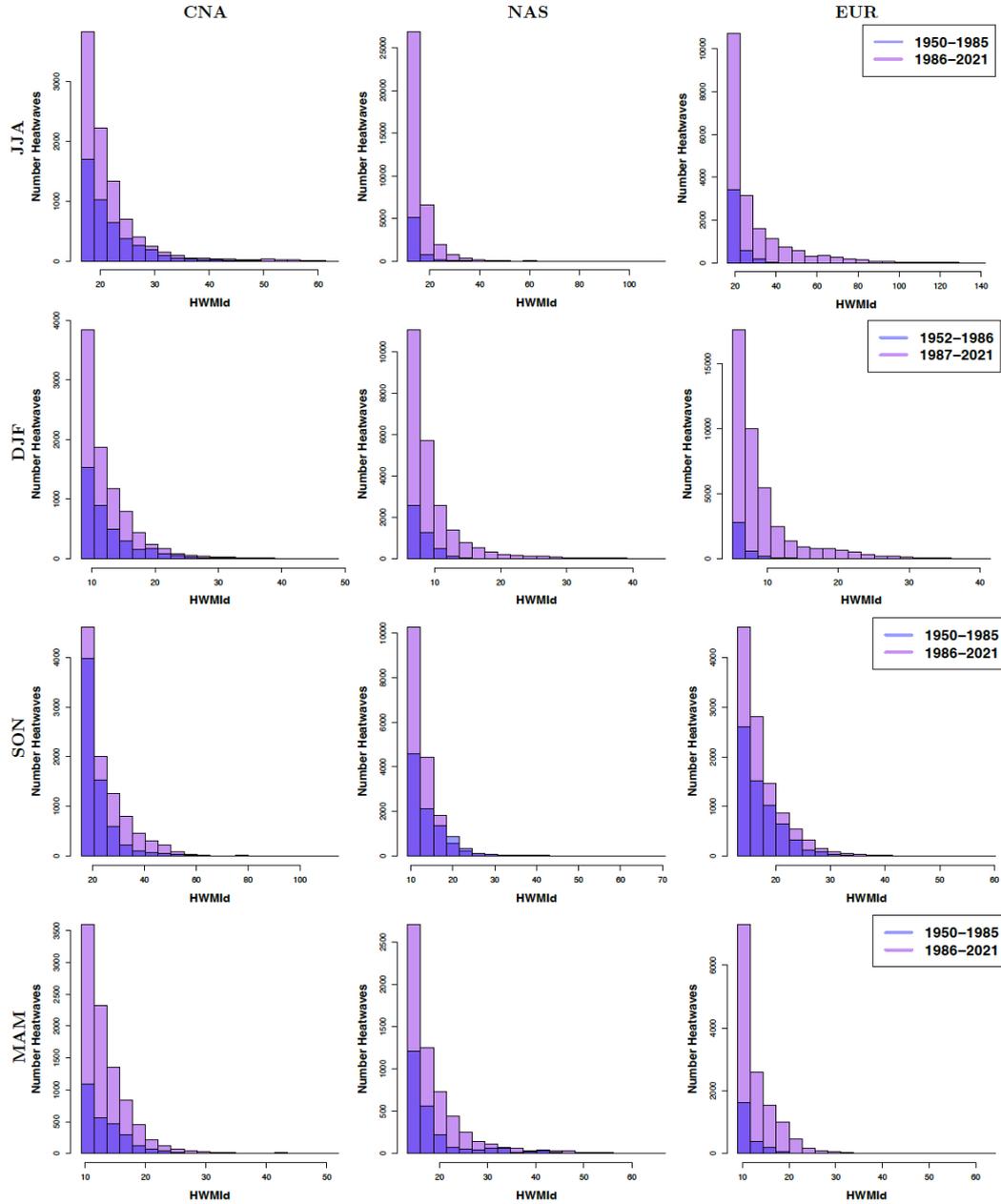


Figure 4. Number of heatwaves with intensities higher than the 99.9th percentile of HWMIId calculated from ERA5-Land over the period 1950-1985, for Central North America (CNA), Northern Asia (NAS) and Northern Europe (NEU). From top to bottom, results for, respectively, JJA, DJF, SON and MAM, are represented through *light blue* bars for the period 1950-1985 and in *purple* for the period 1986-2021. The regions where the bars for both periods overlap are highlighted in darker purple.

311 The presented analyses additionally include, for the first time, an evaluation of ERA5-
 312 Land in terms of the interannual variability of seasonal maxima of daily maximum tem-
 313 peratures against the Berkeley Earth gridded observations and the JRA-55 reanalysis
 314 product, at a global scale. For all seasons, over a large part of the Southern Hemisphere,
 315 ERA5-Land exhibits correlations against the other datasets lower than 0.5, and in some
 316 cases even negative values. The areas for which ERA5-Land is in better agreement with
 317 the other two datasets, in all seasons, are Europe, Central North America and North-
 318 ern Asia.

319 For these three regions where ERA5-Land is in better agreement with the other
 320 datasets, a grid-point-based analysis of the trends of heatwaves over the distinct peri-
 321 ods 1950-1985 and 1986-2021 is performed, considering a cumulative index for heatwave
 322 intensity based on standardized anomalies of maximum daily temperatures. In many of
 323 the considered seasons and regions there is a clear increase in the number and intensity
 324 of heatwaves over the recent years 1986-2021, compared to what was considered very ex-
 325 treme during the period 1950-1985. Europe is the area where the most pronounced changes
 326 between the two periods emerge, with the total number of very extreme events increas-
 327 ing more than ten-fold in boreal winter, and the maximum intensity reaching values up
 328 to three times higher in summer during the most recent times: what was virtually im-
 329 possible during the period 1950-1985 has become more common and extreme in the suc-
 330 cessive 36 years.

331 This study sets the basis for a more unified use of metrics for the calculation of heat-
 332 wave intensity, at the same time providing an analysis of the trends in the number and
 333 intensity of very extreme events over the period from 1950 to 2021, for selected regions,
 334 revealing exceptionally severe changes in heatwaves.

335 5 Open Research

336 5.1 Data Availability Statement

337 The ERA5-Land hourly near surface temperature data used for the computation
 338 of the different heatwave indices proposed in the study are available at the ECMWF Coper-
 339 nicus Climate Change Service (C3S) Climate Data Store (CDS) via [https://doi.org/
 340 10.24381/cds.e2161bac](https://doi.org/10.24381/cds.e2161bac). The Berkeley-Earth gridded daily maximum temperature ob-
 341 servational data are available at <http://berkeleyearth.org/data/>. 6-hourly JRA-55 near-
 342 surface temperature data are available at the Research Data Archive of the National Cen-
 343 ter for Atmospheric Research, Computational and Information Systems Laboratory, via
 344 [ds628.0|DOI:10.5065/D6HH6H41](https://doi.org/10.5065/D6HH6H41).

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351 References

- 352 Breshears, D., Fontaine, J., Ruthrof, K., Field, J., Feng, X., Burger, J., . . . Hardy,
 353 G. (2021). Underappreciated plant vulnerabilities to heat waves. *New Phytolo-*
 354 *gist*, *231*(1), 32–39.
- 355 Cowan, T., Purich, A., Perkins, S., Pezza, A., Boschat, G., & Sadler, K. (2014).
 356 More frequent, longer, and hotter heat waves for australia in the twenty-first
 357 century. *Journal of Climate*, *27*(15), 5851–5871.

- 358 Cueto, R. G., Martínez, A., & Ostos, E. (2010). Heat waves and heat days in an
 359 arid city in the northwest of Mexico: current trends and in climate change
 360 scenarios. *International journal of biometeorology*, *54*(4), 335–345.
- 361 Cusack, L., de Crespigny, C., & Athanasos, P. (2011). Heatwaves and their impact
 362 on people with alcohol, drug and mental health conditions: a discussion pa-
 363 per on clinical practice considerations. *Journal of advanced nursing*, *67*(4),
 364 915–922.
- 365 De Perez, E., Van Aalst, M., Bischiniotis, K., Mason, S., Nissan, H., Pappenberger,
 366 F., ... Van Den Hurk, B. (2018). Global predictability of temperature ex-
 367 tremes. *Environmental Research Letters*, *13*(5), 054017.
- 368 Fischer, E., & Schär, C. (2010). Consistent geographical patterns of changes in high-
 369 impact European heatwaves. *Nature geoscience*, *3*(6), 398–403.
- 370 Forzieri, G., Bianchi, A., Silva, F., Herrera, M., Leblois, A., Lavalle, C., ... Feyen,
 371 L. (2018). Escalating impacts of climate extremes on critical infrastructures in
 372 Europe. *Global environmental change*, *48*, 97–107.
- 373 García-León, D., Casanueva, A., Standardi, G., Burgstall, A., Flouris, A., & Nybo,
 374 L. (2021). Current and projected regional economic impacts of heatwaves in
 375 Europe. *Nature communications*, *12*(1), 1–10.
- 376 Hersbach, H., Bell, B., Berrisford, P., Hirahara, S., Horányi, A., Muñoz-Sabater, J.,
 377 ... others (2020). The ERA5 global reanalysis. *Quarterly Journal of the Royal*
 378 *Meteorological Society*, *146*(730), 1999–2049.
- 379 Holbrook, N., Hernaman, V., Koshiha, S., Lako, J., Kajtar, J., Amosa, P., & Singh,
 380 A. (2022). Impacts of marine heatwaves on tropical western and central Pacific
 381 island nations and their communities. *Global and Planetary Change*, *208*,
 382 103680.
- 383 Iwasaki, A., & Noda, T. (2018). A framework for quantifying the relationship be-
 384 tween intensity and severity of impact of disturbance across types of events
 385 and species. *Scientific reports*, *8*(1), 1–7.
- 386 Kobayashi, S., Ota, Y., Harada, Y., Ebata, A., Moriya, M., Onoda, H., ... Taka-
 387 hashi, K. (2015). The JRA-55 reanalysis: General specifications and basic
 388 characteristics. *Journal of the Meteorological Society of Japan. Ser. II*, *93*(1),
 389 5–48.
- 390 Kovats, R., & Kristie, L. (2006). Heatwaves and public health in Europe. *European*
 391 *journal of public health*, *16*(6), 592–599.
- 392 López-Bueno, J., Navas-Martín, M., Linares, C., Mirón, I., Luna, M., Sánchez-
 393 Martínez, G., ... Díaz, J. (2021). Analysis of the impact of heat waves on
 394 daily mortality in urban and rural areas in Madrid. *Environmental research*,
 395 *195*, 110892.
- 396 Maggiotto, G., Miani, A., Rizzo, E., Castellone, M., & Piscitelli, P. (2021). Heat
 397 waves and adaptation strategies in a Mediterranean urban context. *Environ-*
 398 *mental research*, *197*, 111066.
- 399 Muñoz-Sabater, J., Dutra, E., Agustí-Panareda, A., Albergel, C., Arduini, G., Bal-
 400 samo, G., ... Thépaut, J. (2021). ERA5-Land: A state-of-the-art global re-
 401 analysis dataset for land applications. *Earth System Science Data*, *13*(9),
 402 4349–4383.
- 403 Perkins, S., & Alexander, L. (2013). On the measurement of heat waves. *Journal of*
 404 *climate*, *26*(13), 4500–4517.
- 405 Perkins, S., Alexander, L., & Nairn, J. (2012). Increasing frequency, intensity and
 406 duration of observed global heatwaves and warm spells. *Geophysical Research*
 407 *Letters*, *39*(20).
- 408 Perkins-Kirkpatrick, S., & Lewis, S. (2020). Increasing trends in regional heatwaves.
 409 *Nature communications*, *11*(1), 1–8.
- 410 Peters-Lidard, C., Blackburn, E., Liang, X., & Wood, E. (1998). The effect of soil
 411 thermal conductivity parameterization on surface energy fluxes and tempera-
 412 tures. *Journal of the Atmospheric Sciences*, *55*(7), 1209–1224.

- 413 Rohde, R., Muller, R., Jacobsen, R., Perlmutter, S., Rosenfeld, A., Wurtele, J., ...
 414 Mosher, S. (2013). Berkeley earth temperature averaging process, geoinfor.
 415 geostat.-an overview, 1, 2. *Geoinformatics Geostatistics An Overview*, 1(2),
 416 20–100.
- 417 Russo, S., Dosio, A., Graversen, R., Sillmann, J., Carrao, H., Dunbar, M., ... Vogt,
 418 J. (2014). Magnitude of extreme heat waves in present climate and their pro-
 419 jection in a warming world. *Journal of Geophysical Research: Atmospheres*,
 420 119(22), 12–500.
- 421 Russo, S., Marchese, A. F., Sillmann, J., & Immé, G. (2016). When will unusual
 422 heat waves become normal in a warming africa? *Environmental Research Let-*
 423 *ters*, 11(5), 054016.
- 424 Russo, S., Sillmann, J., & Fischer, E. (2015). Top ten european heatwaves since 1950
 425 and their occurrence in the coming decades. *Environmental Research Letters*,
 426 10(12), 124003.
- 427 Russo, S., & Sterl, A. (2011). Global changes in indices describing moderate temper-
 428 ature extremes from the daily output of a climate model. *Journal of Geophysi-*
 429 *cal Research: Atmospheres*, 116(D3).
- 430 Schaeffer, A., & Roughan, M. (2017). Subsurface intensification of marine heatwaves
 431 off southeastern australia: the role of stratification and local winds. *Geophysi-*
 432 *cal Research Letters*, 44(10), 5025–5033.
- 433 Stone Jr, B., Mallen, E., Rajput, M., Gronlund, C., Broadbent, A., Krayenhoff,
 434 E., ... Georgescu, M. (2021). Compound climate and infrastructure events:
 435 how electrical grid failure alters heat wave risk. *Environmental Science &*
 436 *Technology*, 55(10), 6957–6964.
- 437 Tavakol, A., Rahmani, V., & Harrington, J. (2020). Evaluation of hot temperature
 438 extremes and heat waves in the mississippi river basin. *Atmospheric Research*,
 439 239, 104907.
- 440 Thompson, V., Kennedy-Asser, A., Vosper, E., Lo, Y., Huntingford, C., Andrews,
 441 O., ... Mitchell, D. (2022). The 2021 western north america heat wave among
 442 the most extreme events ever recorded globally. *Science advances*, 8(18),
 443 eabm6860.
- 444 Williams, S., Nitschke, M., Weinstein, P., Pisaniello, D., Parton, K., & Bi, P. (2012).
 445 The impact of summer temperatures and heatwaves on mortality and morbidity
 446 in perth, australia 1994–2008. *Environment international*, 40, 33–38.
- 447 Yu, S., Yan, Z., Freychet, N., & Li, Z. (2020). Trends in summer heatwaves in
 448 central asia from 1917 to 2016: Association with large-scale atmospheric circu-
 449 lation patterns. *International Journal of Climatology*, 40(1), 115–127.