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1 **Does the meteorological origin of heat waves influence their impact on health? A 6-**  
2 **year morbidity and mortality study in Madrid (Spain)**

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21 Abstract.

22 Background: In Spain, two synoptic-scale conditions influence heat wave formation. The first  
23 involves advection of warm and dry air masses carrying dust of Saharan origin (North African  
24 Dust (NAF) = 1). The second entails anticyclonic stagnation with high insolation and stability  
25 (NAF) = 0). Some studies show that the meteorological origin of these heat waves may affect  
26 their impact on morbidity and mortality.

27 Objective: To determine whether the impact of heat waves on health outcomes in Madrid  
28 (Spain) during 2013-2018 varied by synoptic-scale condition.

29 Methodology: Outcome data consist of daily mortality and daily hospital emergency admissions  
30 (morbidity) for natural, circulatory, and respiratory causes. Predictors include daily maximum  
31 and minimum temperatures and daily mean concentrations of  $\text{NO}_2$ ,  $\text{PM}_{10}$ ,  $\text{PM}_{2.5}$ ,  $\text{NO}_2$ , and  $\text{O}_3$ .  
32 Analyses adjust for insolation, relative humidity, and wind speed. Generalized linear models  
33 were performed with Poisson link between the variables controlling for trend, seasonality, and  
34 auto-regression in the series. Relative Risks (RR) and Attributable Risks (AR) were determined.  
35 The RRs for mortality attributable to high temperatures were similar regardless of NAF status.  
36 For hospital admissions, however, the RRs for hot days with NAF=0 are higher than for days with  
37 NAF=1. We also found that atmospheric pollutants worsen morbidity and mortality, especially  
38  $\text{PM}_{10}$  concentrations when NAF=1 and  $\text{O}_3$  concentrations when NAF=0.

39 Results: The effect of heat waves on morbidity and mortality depends on the synoptic situation.  
40 The impact is greater under anticyclonic stagnation conditions than under Saharan dust  
41 advection. Further, the health impact of pollutants such as  $\text{PM}_{10}$  and  $\text{O}_3$  varies according to the  
42 synoptic situation.

43 Conclusions: Based on these findings, we strongly recommend prevention plans to include data  
44 on the meteorological situation originating the heat wave, on a synoptic-scale, as well as  
45 comprehensive preventive measures against the compounding effect of high temperatures and  
46 pollution.

47

48

49 1. Introduction

50 According to the latest Intergovernmental Panel on Climate Change (IPCC) report on the  
51 impacts, adaptations and vulnerabilities stemmed from climate change (IPCC 2022), heat waves  
52 will, undoubtedly, increase in frequency and intensity. Thus, save for a progressive adaptation  
53 to higher temperatures, their health impact will only worsen (Díaz et al., 2019).

54 In many countries, the heat-related impact on health has decreased significantly in recent  
55 decades (Schifano et al., 2012; Chung et al., 2017; Díaz et al., 2018a, Sheridan & Dixon, 2017).  
56 The reasons are multifactorial, including geographical variability (high temperatures have less  
57 impact on health the warmer the location, probably due to higher heat habituation). This leads  
58 to greater use of air conditioning, better health services, and improvements in insulation in  
59 housing and in infrastructures in general, among others (Martínez GS et al., 2019). However,  
60 beyond any doubt, a key factor in this observed decrease in impact is the establishment of  
61 prevention plans (WHO, 2018) in 66% of European countries.

62 Improvements made to these plans would undoubtedly benefit individuals' health on  
63 particularly hot days. Some of these improvements are epidemiological in nature, i.e., they  
64 determine at which temperature prevention plans should be implemented based on  
65 epidemiological temperature-mortality studies, rather than relying solely on climatic indices  
66 (Andersen et al., 2021). Additional improvements are based on the different meteorological  
67 patterns on a synoptic scale that condition the atmosphere favoring the high temperatures  
68 characteristic of heat waves (Yoon et al., 2018; Sfiică et al., 2017). Previous studies conclude that  
69 the severity of dangerous heat waves is directly related to the specific meteorological conditions  
70 originating them (Kalkstein et al 2011; Hajat et al., 2010; Metzger et al., 2010). Clearly, this is in  
71 addition to the usual factors typifying a heat wave such as intensity and duration (Diaz et al.,  
72 2002).

73 Thus, data on synoptic meteorological conditions originating heat waves have been increasingly  
74 considered in research starting with Kalkstein and Greene for Central and Eastern U.S. (Kalkstein  
75 & Green, 1997) and later redefined by Sheridan and Kalkstein for Canada and Western U.S.  
76 (Sheridan & Kalkstein, 2004) as well as Bower and colleagues for Western Europe (Bower et al.,  
77 2007).

78 Synoptic-scale meteorological conditions present during heat waves have been analyzed for  
79 Spain (Serrano-Notivoli et al., 2022). Most cases involve strong anticyclonic stagnation  
80 conditions produced by the Azores anticyclone in the absence of wind and with high insolation  
81 levels. This stagnation situation, which by itself can generate a heat wave, can be amplified by

82 the advection of warm and dry air from North Africa and particulate matter from the Sahara.  
83 The result is a more intense and longer heat wave compared to those resulting solely from an  
84 anticyclonic blockade (Serrano-Notivoli et al., 2022).

85 The aim of this study is to analyze whether the impact of high temperatures on morbidity and  
86 mortality from all natural causes and from certain specific causes is modified by the synoptic-  
87 scale meteorological event generating those temperatures. Specifically, we differentiate  
88 between heat waves generated by an advection of dust from the Sahara and those created  
89 exclusively as the result of a situation of anticyclonic stagnation. In this study we analyze data  
90 for the province of Madrid (Spain) (2022 population: 6.7 million) with the idea of including all  
91 the 17 Spanish regions (total population: 47.5 million) in future analyses. The current High  
92 Temperature Prevention Plan of the Spanish Ministry of Health (Ministry of Health, 2022) does  
93 not take into account the meteorological origin of heat waves to calculate the impact of  
94 temperatures on mortality. The results of this study aim to propose to the health authorities  
95 that subsequent updates of the Plan take into account the different origin of heat waves to  
96 calculate the impact of temperature and air pollution on daily mortality.

## 97 2. Materials and Methods.

### 98 2.1. Direct Variables

99 Independent variables include six years worth of meteorological and air pollution data recorded  
100 between January 1<sup>st</sup>, 2013 and December 31<sup>st</sup>, 2018. The meteorological data were collected in  
101 the meteorological observatory of reference located in the district of Retiro in the downtown  
102 area of the city of Madrid. It was specifically chosen because it provided the daily maximum  
103 temperature data used to determine the official threshold temperature defining a heat wave for  
104 the Community of Madrid according to the Spanish Ministry of Health (Ministerio de Sanidad,  
105 2022). The meteorological data examined were: Daily maximum and minimum temperatures  
106 (Tmax and Tmin, respectively), average values in Celsius (°C); daily average wind speed (km/h);  
107 daily insolation or sunlight hours (hours) and daily average relative humidity (%). These data  
108 came from the State Meteorological Agency (AEMET for its Spanish acronym).

109 Pollution data correspond to the average daily concentrations of the pollutants PM<sub>10</sub>, PM<sub>2.5</sub>, NO<sub>2</sub>,  
110 and O<sub>3</sub> (all in µg/m<sup>3</sup>). These represent the averages of the mean concentration values from every  
111 meteorological station located in the Community of Madrid. These data were provided by the  
112 Ministry for Ecological Transition and Demographic Challenge (MITERD for its Spanish acronym).

113 Based on data provided by the Spanish National Institute of Statistics (INE for its Spanish  
114 acronym) we examined six outcome variables: daily mortality (3 causes) and morbidity (3  
115 causes). The mortality variable consisted on the average daily mortality reported in  
116 municipalities with populations over 10,000 inhabitants in the Community of Madrid between  
117 2013 and 2018. We included mortality for all natural causes (ICD-10: A00-R99), circulatory  
118 causes (ICD-10: I00-I99), and respiratory causes (ICD-10: J00-J99). As a measure of heat wave-  
119 related morbidity, we used daily emergency hospital admissions based on the INE's annual  
120 Hospital Morbidity Survey data. Specifically, we analyzed daily-unscheduled emergency hospital  
121 admissions during the study period for the same causes and ICD-10 codes as mentioned above.

## 122 2.2. Derived Variables.

123 We recoded the variables above to create additional variables reflecting different actual  
124 functional relationships among dependent and independent variables.

125 To account for the impact of high temperatures on morbidity and mortality, we adopted the  
126 definition of a "heat wave" used by the Spanish Ministry of Health for the Community of  
127 Madrid, i.e., a daily  $T_{\max}$  of 34 °C. We justify the use of  $T_{\max}$ , rather than  $T_{\min}$ , based on the  
128 results reported by different studies indicating that it is the daily  $T_{\max}$ , which actually better  
129 correlates with mortality during heat waves (Guo et al., 2017; Alberdi et al., 1998; Díaz et al.,  
130 2002).

131 Thus, heat wave is defined by the variable  $T_{\text{heat}}$  as shown below (Díaz et al., 2015):

$$132 \quad T_{\text{heat}} = 0 \quad \text{if } T_{\max} < 34^{\circ}\text{C}$$

$$133 \quad T_{\text{heat}} = T_{\max} - 34 \quad \text{if } T_{\max} \geq 34^{\circ}\text{C}$$

134 From a health point of view, if in a single day  $T_{\max}$  surpass 34°C, already has an impact on  
135 health; it will be considered as heat wave. The concept of heat wave refers to one or more  
136 consecutive hot days, with the number of such days termed the heat wave's duration (Díaz et  
137 al., 2002; Montero et al., 2010; Kent et al., 2014; Guo et al., 2017; Kang et al., 2020). The higher  
138 the  $T_{\text{heat}}$  values, the greater the intensity of the heat wave.

139 For the pollutants analyzed, we assume a linear relationship with morbidity and mortality with  
140 no threshold for  $\text{PM}_{10}$ ,  $\text{PM}_{2.5}$  (Ortiz et al., 2017; Arroyo et al., 2019a; Reyes et al., 2014; Samoli  
141 et al., 2014; Sera et al., 2019), and  $\text{NO}_2$  (Linares et al., 2018; Meng et al., 2021; He et al., 2020;  
142 Arroyo et al., 2019b). In the case of ozone ( $\text{O}_3$ ), we assume a quadratic relationship with daily  
143 morbidity and mortality (Zang et al., 2021; Bell et al., 2006; Malley et al., 2017; Maté et al., 2010).

144 Previous studies in Spain show that the threshold value for a negative health impact for daily  
145 average ozone concentrations for the Community of Madrid is set at 60  $\mu\text{g}/\text{m}^3$  (Díaz et al.,  
146 2018b). Thus, we created a new variable,  $O_{3h}$ , defined as follows:

$$147 \quad O_{3h} = 0 \quad \text{if } O_3 < 60 \mu\text{g}/\text{m}^3$$

$$148 \quad O_{3h} = O_3 - 60 \quad \text{if } O_3 \geq 60 \mu\text{g}/\text{m}^3$$

149 However, the effect of the independent variables on daily mortality and morbidity levels may  
150 come about on the same day or with a time lag. For heat, lags of up to 5 days have been included  
151 (Díaz et al; 2002). For  $\text{PM}_{10}$ ,  $\text{PM}_{2.5}$ , and  $\text{NO}_2$  concentrations we included up to a 5-day lag (Ortiz  
152 et al., 2016, Linares et al., 2018), and for ozone concentrations up to a 9-day lag (Díaz et al.,  
153 2018b) were included. For the rest of the meteorological variables, and since no previous studies  
154 have included them simultaneously, we considered time lags of up to 14 days.

155 As mentioned earlier in the introduction, the meteorological patterns on a synoptic-scale  
156 associated to heat waves in the Community of Madrid are connected to the position of the  
157 Azores anticyclone, by itself capable of producing heat waves. These are often intensified by the  
158 intrusion of very warm North Africa winds carrying suspended dust from the Sahara (Serrano-  
159 Notivoli et al., 2022). Therefore, from a meteorological point of view and for the purpose of this  
160 paper, we classify heat waves into two categories:

161 1. A heat wave is classified as North African Dust (NAF)=1 when advection of Saharan  
162 dust is detected.

163 2. A heat wave is classified as NAF=0 when caused by an anticyclonic stagnation  
164 triggered by the Azores anticyclone, and characterized by strong insolation but hardly  
165 any wind.

166

167 During the study period, days are classified as NAF=1 when, according to information provided  
168 by MITERD by region (MITECO, 2019), Saharan dust advections are detected in the Central  
169 region of Spain. All other days during a heat wave will be classified as non-dust advection, i.e.,  
170 NAF=0.

### 171 2.3. Other control variables.

172 The dependent and independent variables may have the same seasonal components, the same  
173 trend or the same autoregressive character (history of the series at short term) which may mean

174 that the significant associations between them are due to these similar components, it is  
175 necessary to include them in the models as control variables.

176 In addition to the independent variables described above with their corresponding time lags,  
177 other variables that influence the trend and seasonality of the series are also taken into account.  
178 For this purpose, a variable  $n_1$  is included. This variable is equal to 1 on the first day of the series,  
179 2 on the second, and so on. The annual, half-yearly, quarterly and bimonthly seasonalities are  
180 controlled by including sine and cosine functions with the aforementioned periods. Likewise,  
181 the days of the week, from Monday to Sunday, and public holidays within the study period will  
182 be considered. Finally, the autoregressive nature of the series will be controlled for with the  
183 inclusion of an autoregressive component of order 1.

184 In the model described below, it can be seen how these variables form part of the models  
185 relating the dependent variables to the independent variables.

#### 186 2.4. Modeling process and calculation of deaths and hospital admissions attributable to heat 187 waves.

188 For each of the six dependent variables and for the two possible heat wave situations (NAF=1  
189 vs. NAF=0), generalized linear models (GLM) with Poisson link were performed. In these models,  
190 all the independent and control variables, as well as the transformed variables, with their  
191 corresponding lags, were introduced.

192 This type of model has been used very frequently when relating morbidity and mortality to  
193 extreme temperatures in time series analysis (Díaz et al., 2015; Carmona et al., 2016a, López-  
194 Bueno et al., 2021; Culqui et al., 2022).

195 These models followed the general formula:

$$196 \log(TM) = a + \beta_1 \text{lag}(TM, 1)_i + \beta_2 n_1 + \beta_3 \text{Seas}_{ij} + \beta_4 \text{lag}(Theat, g) + \beta_5 \text{lag}(PM10, h) \\ 197 + \beta_6 \text{lag}(O3h, l) + \beta_k \text{lag}(\text{Other } k, m)$$

198 Where TM represents the mortality rate,  $a$  is the intercept,  $\beta_k$  represents the coefficient of  
199 each of the  $k$  variables,  $\text{lag}(TM, 1)$  is the autoregressive component of the first order of mortality  
200 in observation  $i$ ;  $n_1$  is the value of the trend in observation  $i$ ;  $\text{Seas}$  represents seasonality  $j$   
201 observation  $i$ ;  $(Theat, g)$  represents  $Theat$  lagged from 0 to 5;  $\text{lag}(PM10, h)$  represents PM10  
202 lagged from 0 to 5;  $\text{lag}(O3h, l)$  represents O3h lagged from 0 to 8;  $\text{lag}(\text{Other } k, m)$  represents  
203 the meteorological and pollution variable  $k$  lagged from 0 to  $m$ .

204 We used a stepwise process to eliminate variables failing to reach statistical significance. Thus,  
205 the final model includes only those variables that were statistically significant at  $p < 0.05$ .

206 From the estimate ( $\beta$ ) of each significant variable and its corresponding confidence interval, the  
207 corresponding Relative Risk (RR) was calculated as  $RR = e^{\beta}$ . These RRs were calculated for  
208 increments of  $T_{\text{heat}}$  of  $1^{\circ}\text{C}$  and increments of  $10\ \mu\text{g}/\text{m}^3$  for the pollutants.

209 Based on the RR values, we calculated the corresponding Attributable Risks (AR) following the  
210 Coste & Spira equation:  $AR = (RR - 1) * 100 / RR$  (Coste & Spira 1991). Based on the AR values, and  
211 following the methodology outlined by Carmona and colleagues (Carmona et al., 2016b), the  
212 number of attributable deaths and hospital admissions were calculated for each variable that  
213 was significant in the modeling process.

214 Data management and analyses were performed using the R software version 4.0.2 and STATA  
215 BE-Basic Edition version 17, IBM SPSS Statistics version 27, and Excel (with the Power Query  
216 editor) from the Microsoft Office Professional Plus 2019 package.

217

218

219

220 Results.

221 Table 1 shows the distribution of the independent variables during heat wave days with NAF  
222 values of 1 or 0 for the period of interest. Values for the primary pollutants (PM<sub>10</sub> and NO<sub>2</sub>) and  
223 for T<sub>max</sub> and T<sub>min</sub> are statistically significantly higher on days with NAF=1 compared to days with  
224 no Saharan dust advection (NAF=0).

225 There were 39 heat waves shaped by NAF=1 meteorological conditions versus 33 generated by  
226 an anticyclonic stagnation pattern (NAF=0). Heat waves based on NAF=1 conditions were longer  
227 and more intense than heat waves due to NAF=0 conditions. Differences were statistically  
228 significant.

229 Table 2 shows the distribution of daily mortality and daily emergency hospital admissions for  
230 natural, circulatory, and respiratory causes across heat wave days with NAF=1 and NAF=0  
231 conditions. Both daily mortality and admissions are higher on days with heat wave and advection  
232 (NAF=1) than on days with heat wave but no Saharan dust advection. All differences detected  
233 are statistically significant except for daily morbidity and mortality due to circulatory causes.

234 Figure 1 and figure 2 show the statistically significant results of the Poisson regression models  
235 for the mortality and hospital admissions outcomes, respectively, according to NAF conditions,  
236 with their corresponding ARs for each day of the heat waves.

237 *As an example, the final model for circulatory mortality on days with NAF = 0 is shown:*

$$\begin{aligned} 238 \log(\text{circ}) &= 3.60 - 0.0037 \log(\text{circ}, 1) - 0.1753 \sin 180 + 0.0601 \log(\text{Theat}, 0) + 0.0043 \log \\ 239 &(\text{O3h}, 2) - 0.00467 \log(\text{wind}, 4). \end{aligned}$$

240 Figure 3 and Figure 4 shows the RRs for the statistically significant variables associated with  
241 those same outcomes. Heat wave days with no Saharan dust advection (NAF=0) have a greater  
242 adverse impact on mortality due to natural causes than days with advection (NAF=1), though  
243 the difference fails to reach statistical significance. The same is true for mortality due to  
244 circulatory causes. In this case, however, there are two other results worth mentioning. First,  
245 the impact of PM<sub>10</sub> concentrations on mortality due to circulatory causes, though only in the  
246 presence of dust advection (NAF=1); and, second, the impact of O<sub>3a</sub> concentrations on that same  
247 outcome, though only on days with no dust advection (NAF=0). Regarding mortality due to  
248 respiratory causes, T<sub>heat</sub> is only a prognostic factor on days with advection of Saharan dust  
249 (NAF=1). Finally, regardless of dust advection conditions, tropospheric ozone has an impact on  
250 death due to respiratory causes.

251 Despite these observed differences, if we add the daily mortality caused by heat and pollutants,  
252 the values are similar on days with or without dust advection. In fact, no differences reach  
253 statistical significance.

254 Regarding results from the Poisson models for daily hospital admissions (Figure 2 and Figure 3),  
255 we find the absence of impact of  $T_{\text{heat}}$  on morbidity on heat wave days with NAF=1 conditions,  
256 remarkable. Especially since heat wave days with NAF=0 conditions do impact hospital  
257 admissions. The increase of hospital admissions on heat wave days with advection (NAF=1)  
258 would be related to  $PM_{10}$  concentrations in all-cause admissions, and to levels of the pollutant  
259 ozone in the case of respiratory-cause admissions. Whereas increases on hospital admissions on  
260 heat wave days due to an anticyclonic stagnation pattern (NAF=0) would be associated to rises  
261 in ozone concentrations. Finally, for both mortality and morbidity, the impact of air pollution on  
262 circulatory-related causes transpires quicker, 0 to 3-day lags, than on respiratory-related causes,  
263 which register 6 to 8-day lags.

264

### 265 3. Discussion.

266 During the six-year period of interest (2013-2018), Spain registered 232 heat wave days. Saharan  
267 dust advections took place in 144 (62.1%) of those days, which were distributed across 39 heat  
268 waves. The other 88 days (37.9%) were distributed across 33 heat waves with anticyclonic  
269 stagnation conditions.

270 Consistent with their meteorological origin, heat waves associated to dust advections reach  
271 higher extreme temperatures ( $T_{\text{heat}}$  values) and longer durations than those related to  
272 anticyclonic conditions only. Although in Madrid heat waves usually stem from an anticyclonic  
273 stagnation pattern, the advection of dust carried by Saharan air flow intensifies the heat waves  
274 effects (Serrano-Notivol et al., 2022).

275 We also observed that for all pollutants, except ozone, their concentrations during heat waves  
276 are statistically significantly higher in dust advection conditions than in anticyclonic stagnations  
277 patterns. This increase is especially striking for  $PM_{10}$ . These results support previous work carried  
278 out in Spain (Moreira et al., 2020), Barcelona (Spain) (Pandolfi et al 2014) and Madrid (Spain)  
279 (Salvador et al., 2019). The increase in concentration of all pollutants, especially  $PM_{10}$ , observed  
280 in heat waves with advection of particulate matter, may be related to a decrease in incident  
281 solar radiation caused by the blocking effect of the suspended particles themselves. Lower  
282 radiation, in turn, may cause convective currents to decrease, which would diminish the

283 thickness of the mixing layer and result in higher pollutant concentrations (Li et al., 2017). This  
284 decrease in solar radiation during dust advection days was also observed in our analyses (Table  
285 1). It is also likely that higher solar radiation on non-dust advection days translates into the ozone  
286 levels not being as low as on dust-advection days, which would render the difference in ozone  
287 levels across the two types of heat wave days not statistically significant.

288 Given the greater intensity and duration of heat waves associated with dust advection, one may  
289 expect high temperatures to have a greater impact on mortality and morbidity in the presence  
290 of suspended dust particles than in their absence; however, our results suggest the opposite.  
291 The impact of both types of heat waves on mortality is very similar. However, their impact on  
292 hospital admissions varies. No impact was observed during dust-advection days but the impact  
293 during heat waves caused by anticyclonic stagnation conditions was quite significant. Therefore,  
294 from the public health perspective, we should avoid classifying all heat waves as having similar  
295 health impacts or risk levels. In fact, heat waves vary significantly in risk level and, furthermore,  
296 shorter and milder heat waves may turn out to be more fatal than longer, more intense ones.  
297 Our findings confirm reports from previous studies (Kalkstein et al. 2011; Hajat et al., 2010;  
298 Metzger et al., 2010). The conclusions of this body of work call for the inclusion of the synoptic  
299 conditions causing each heat wave as part of the data informing health-related prevention plans  
300 for high temperatures (Kalkstein & Greene, 1997; Sheridan & Kalkstein 2004; Bower et al., 2007;  
301 Zhang et al., 2012).

302 The lack of association between the presence of extreme temperatures and morbidity in heat  
303 wave days with Saharan dust advection may seem to suggest that the strong impact on health  
304 related to these very high temperatures causes immediate death and, thus, the individual is not  
305 even admitted to the hospital (Linares & Díaz, 2008; Mastrangelo et al., 2006); thus, not  
306 impacting morbidity. However, our results do not support this hypothesis since the relative risks  
307 for mortality during heat wave days with dust advection are not higher than the relative risks  
308 during heat wave days with no dust advection.

309 One possible explanation for this fact could be that the first heat waves of each year normally  
310 were originated in situations of anticyclonic blocking. This is the situation analyzed in this study  
311 and these first heat waves have the greatest effect on mortality due to the greater number of  
312 people susceptible to heat (Díaz et al., 2002). Furthermore, as explained in the introduction,  
313 heat waves in Spain usually start with a situation that can be amplified by the advection of warm  
314 and dry air from North Africa and particulate matter from the Sahara (Serrano-Notivoli et al.,  
315 2022). These linked events entail a greater effect on mortality of the heat waves at the beginning

316 of each wave (Díaz et al., 2002) and, therefore, a greater impact due to situations of anticyclonic  
317 blocking.

318 Our analyses also show that, in addition to the health impact of intense heat, the impact of  
319 pollutants on both daily hospital admissions and mortality is not only notable but greater than  
320 the impact of very high temperatures. This impact also varies by type of heat wave. On heat  
321 wave days with dust advection the impact of PM<sub>10</sub> on health outcomes is predominant, whereas  
322 in the absence of dust advection the only pollutant with a significant health impact is  
323 tropospheric ozone. These observations confirm findings from similar studies conducted in Spain  
324 and elsewhere in Europe regarding suspended Saharan dust and mortality (Díaz et al., 2017;  
325 Stafoggia et al., 2016) and morbidity (Reyes et al., 2014). Therefore, the health consequences of  
326 any heat wave are not only related to the number of days with temperatures above 34°C, but  
327 also to pollutants acting synergistically. In sum, high temperatures and pollution may boost the  
328 impact of both PM<sub>10</sub> (Parry et al., 2019) and ozone (Yang et al., 2022).

329 Conventionally, heat wave prevention plans focus exclusively on temperature-related effects.  
330 Our results strongly suggest that these plans must be more comprehensive (Linares et al., 2020),  
331 i.e., they should integrate all factors with potential health impacts that may be exacerbated by  
332 a heat wave. These include the aforementioned increase in air pollution, forest fires (Linares et  
333 al., 2018b), the increase in foodborne diseases (Duchenne-Moutien & Neetoo, 2021), and the  
334 exacerbation of droughts (Salvador et al., 2020).

### 335 3.1. Limitations of the study.

336 We followed the methodology commonly used in this type of studies (Samet et al., 2000; Díaz  
337 et al., 2015; Linares et al., 2018a, b; López-Bueno et al., 2021). We have tried to minimize any  
338 potential methodological biases by including in our models all relevant control variables  
339 available in our data such as seasonality, trend, days of the week, vacation periods, and  
340 autoregressive nature of the series.

341 As an ecological study, there are additional limitations such as the difficulty of extrapolating our  
342 results, applicability to the general population, to the individual level. In addition, there are  
343 limitations inherent to the representativeness of the exposure of each individual to the  
344 environmental variables considered (Barceló et al., 2016). Although the network of weather  
345 stations collecting air pollution data is very extensive, working with average concentrations  
346 could introduce a bias in the results. Further, data for all meteorological variables were collected  
347 in a single observatory, which may also bias our results, despite this being the observatory of  
348 reference of the Madrid region (Díaz et al., 2002). No specific validation was performed within

349 the project to assess representativeness of spatial variability in air pollutants, thus, our study  
350 suffers from Berkson-type measurement error (Barceló et al., 2016). In addition, the inevitable  
351 misclassification of the causes of hospital admissions also introduces some errors.

352 Finally, it should be noted that the data for this study came from only one province in one of the  
353 9 regions of interest. Whereas Spain is geographically and politically divided into 17 autonomous  
354 regions, for the study of Saharan dust advections, Spain is divided into 9 regions (MITECO, 2019),  
355 so it would be necessary to extend it to at least one province for each of these 9 regions.

#### 356 4. Conclusions.

357 Our findings indicate that heat waves originating in anticyclonic stagnation patterns have a  
358 greater impact on morbidity (measured here as daily hospital admissions) than heat waves  
359 characterized by Saharan dust advections. This is so despite the fact that the latter tend to be  
360 more intense and last longer periods. Thus, prevention plans should take the synoptic-scale  
361 meteorological origin of the heat wave into account in order to be more effective. In addition,  
362 on heat wave days the concentration of the pollutants PM<sub>10</sub> and ozone undergo important  
363 increases, which have an even greater impact on mortality and morbidity than the very high  
364 temperatures. Therefore, prevention plans should include both risk factors, type of heat wave  
365 and pollutant levels, in their estimates to improve their implementation and effectiveness.

#### 366 Disclaimer

367 The researchers declare that they have no conflict of interest that would compromise the  
368 independence of this research work. The views expressed by the authors are not necessarily  
369 those of the institutions they are affiliated with.

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563

**Table 1. Descriptive statistics of the independent variables on heat wave days with and without Saharan dust advection.**

	Dust advection (NAF <sup>a</sup> = 1) N= 144					No dust advection (NAF= 0) N= 88				
	Mean	Max	Min	SD <sup>b</sup>	CV <sup>c</sup>	Mean	Max	Min	SD <sup>b</sup>	CV <sup>c</sup>
<b>PM<sub>10</sub> (µg/m<sup>3</sup>)*</b>	33.6	85.7	16.5	10.6	<b>32</b>	22.2	32.2	12.3	4.4	<b>20</b>
<b>PM<sub>2.5</sub> (µg/m<sup>3</sup>)*</b>	14.9	33.1	6.7	3.9	<b>26</b>	11	17.9	6.1	2.5	<b>23</b>
<b>NO<sub>2</sub> (µg/m<sup>3</sup>)*</b>	28.2	51.8	12.8	8.4	<b>30</b>	24.5	47.8	11.1	6.2	<b>25</b>
<b>O<sub>3</sub> (µg/m<sup>3</sup>)</b>	83.4	112.1	47.4	13.2	<b>16</b>	80.5	113.7	47.5	12.6	<b>16</b>
<b>T<sub>max</sub><sup>d</sup> (°C)*</b>	36.2	40	34.1	1.6	<b>4</b>	35.6	39.2	34.1	1.2	<b>3</b>
<b>T<sub>min</sub><sup>e</sup> (°C)*</b>	22.2	25.9	17	1.7	<b>8</b>	21.4	25.1	17.9	1.3	<b>6</b>
<b>Wind speed (km/h)</b>	6.4	10.5	2.8	1.5	<b>23</b>	6.6	10.7	3.2	1.4	<b>21</b>
<b>Insolation (hours)</b>	12.1	14.4	2.1	1.9	<b>16</b>	12.9	14.4	7.9	1.4	<b>11</b>
<b>Relative humidity (%)</b>	39.5	61.9	28.9	5.6	<b>14</b>	40.4	53.8	30.6	4.5	<b>11</b>
<b>T<sub>heat</sub><sup>f</sup> (°C)*</b>	2.2	6	0.1	1.6	<b>73</b>	1.6	5.2	0.1	1.3	<b>81</b>
<b>Heat wave duration (days)*</b>	3.7	15	1	3	<b>81</b>	2.7	7	1	1.6	<b>59</b>

\* Statistically significant differences at p<0.05.

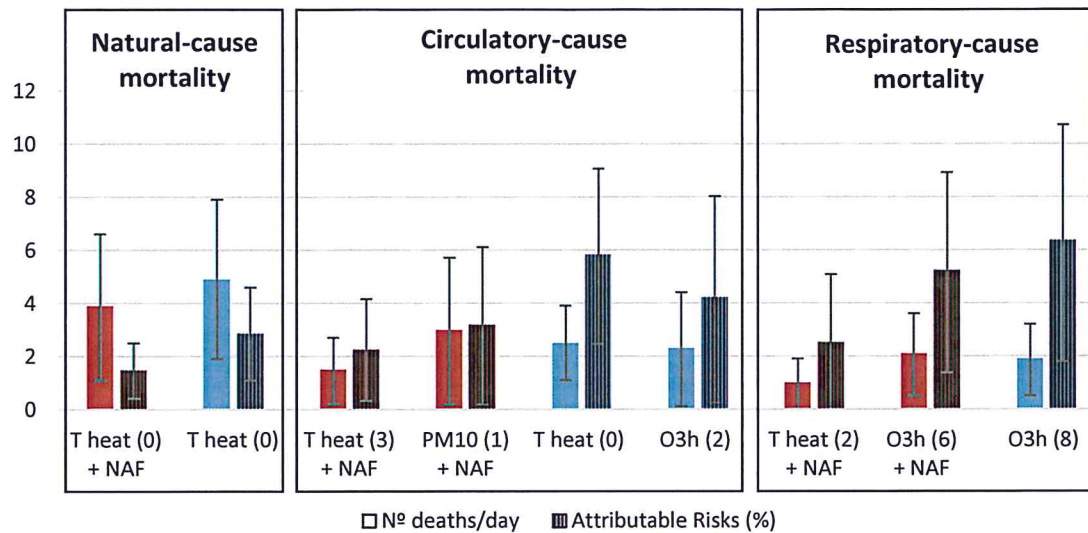
<sup>a</sup> North African Dust; <sup>b</sup> Standard Deviation; <sup>c</sup> Coefficient of variation %; <sup>d</sup> Daily maximum temperature; <sup>e</sup> Daily minimum temperature; <sup>f</sup> Degrees of daily temperature in excess above 34°C.

**Table 2. Descriptive statistics of cause-specific daily mortality and daily hospital admissions according to the presence of Saharan dust advection on heat wave days.**

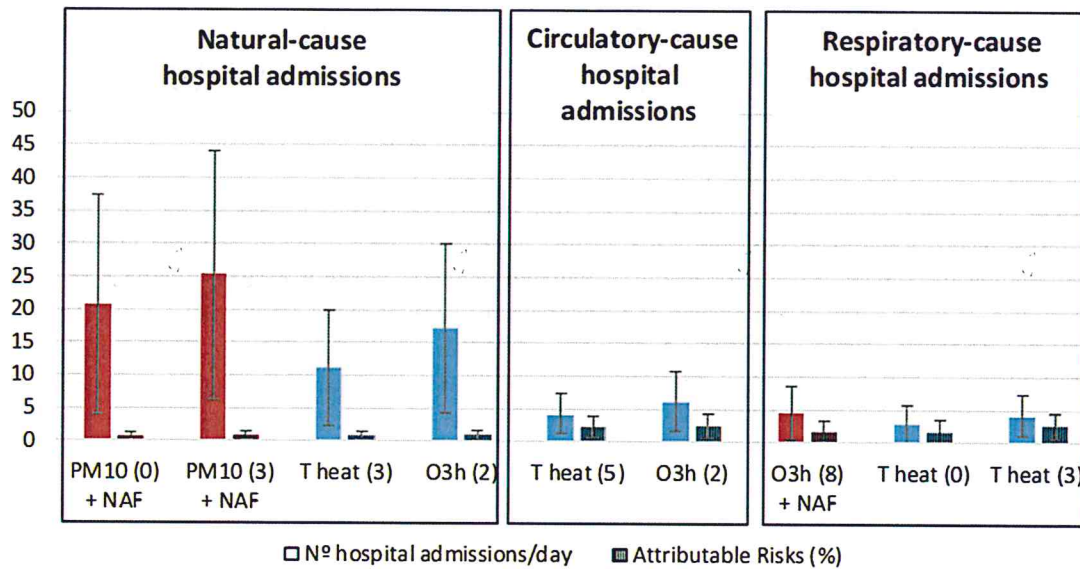
	<b>Dust advection (NAF<sup>a</sup>= 1) N= 144</b>					<b>No dust advection (NAF<sup>a</sup>= 0) N= 88</b>				
	<b>Mean</b>	<b>Max</b>	<b>Min</b>	<b>SD<sup>b</sup></b>	<b>CV<sup>c</sup></b>	<b>Mean</b>	<b>Max</b>	<b>Min</b>	<b>SD<sup>b</sup></b>	<b>CV<sup>c</sup></b>
<b>Natural causes mortality*</b>	113.7	168	78	18.3	<b>16</b>	105.8	135	72	13.7	<b>13</b>
<b>Circulatory causes mortality</b>	27.7	45	12	6.5	<b>23</b>	26	42	13	6.4	<b>25</b>
<b>Respiratory causes mortality*</b>	16.1	34	6	5.2	<b>32</b>	14.4	23	5	3.7	<b>27</b>
<b>Natural causes admissions*</b>	864.4	1131	537	136.6	<b>16</b>	824.3	1034	545	123.4	<b>15</b>
<b>Circulatory causes admissions</b>	124.8	194	64	26.3	<b>21</b>	119.9	171	54	26.7	<b>22</b>
<b>Respiratory causes admissions*</b>	112.1	197	52	28.3	<b>25</b>	100.8	161	50	22.7	<b>23</b>

\*Statistically significant differences at  $p < 0.05$ .

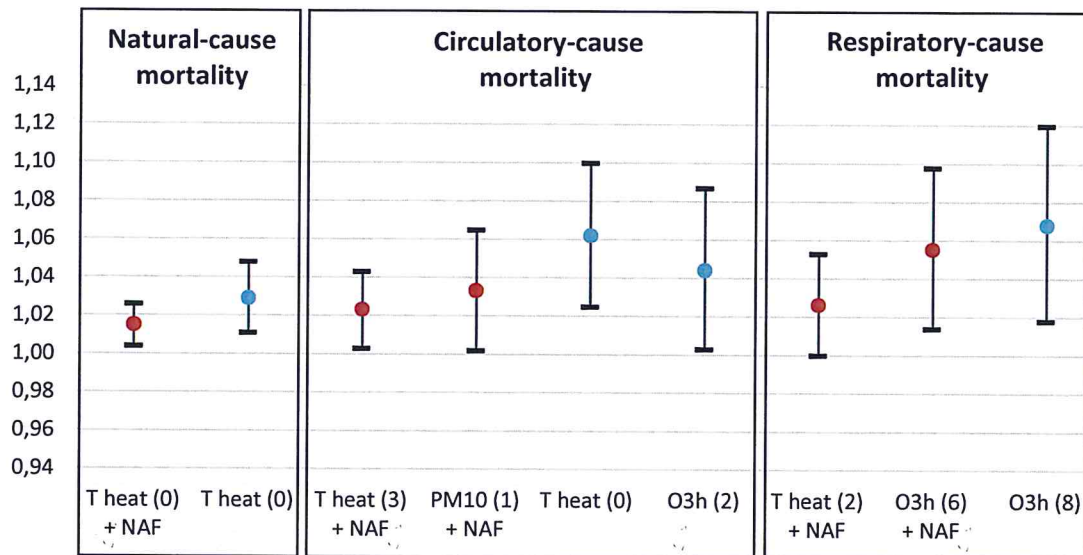
<sup>a</sup> North African Dust; <sup>b</sup> Standard Deviation; <sup>c</sup> Coefficient of variation %.



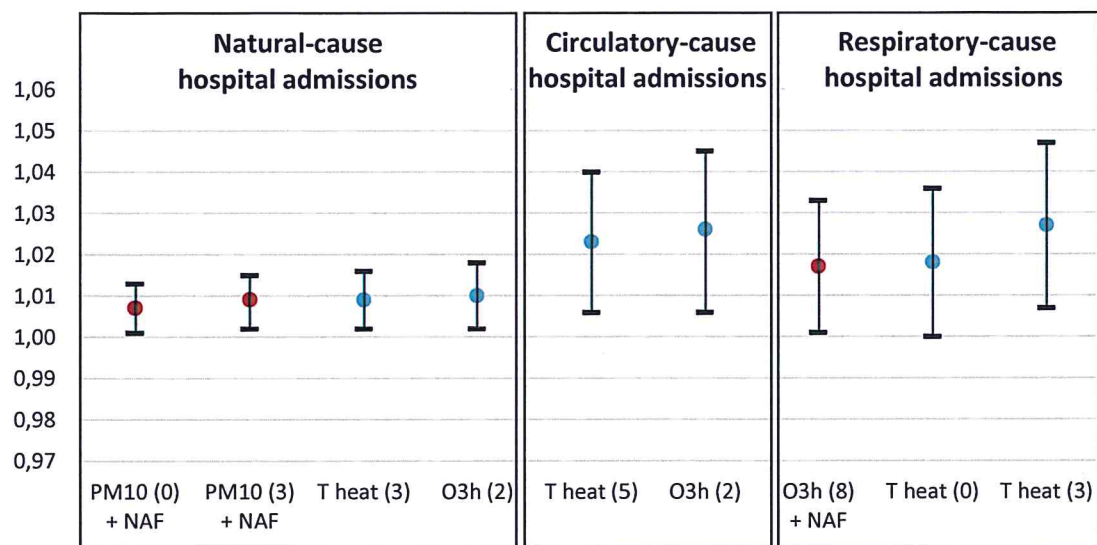
**Figure 1.** Statistically significant variables derived from Poisson models for cause-specific daily mortality. In red: values corresponding to days with Saharan dust advection (NAF: North African Dust). In blue: values corresponding to days with no Saharan dust advection (NAF=0). Without stripes: numbers of deaths attributable to each variable for each day of heat wave. With stripes: Attributable Risks (AR) in % corresponding to 10 g/m<sup>3</sup> increase in pollutants and 1 °C increase in T<sub>heat</sub>. AR and numbers of deaths attributable to each variable for each day of heat wave with 95% Confidence Interval. T heat: degrees of daily temperature in excess of 34 °C. O3h: ozone high. In parentheses: time lag, in days, in which the association occurs.



**Figure 2.** Statistically significant variables derived from Poisson models for cause-specific daily hospital admissions. In red: values corresponding to days with Saharan dust advection (NAF: North African Dust). In blue: values corresponding to days with no Saharan dust advection (NAF=0). Without stripes: numbers of hospital admissions attributable to each variable for each day of heat wave. With stripes: Attributable Risks (AR) in % corresponding to 10 g/m<sup>3</sup> increase in pollutants and 1 °C increase in T<sub>heat</sub>. AR and numbers of hospital admissions to each variable for each day of heat wave with 95% Confidence Interval. T heat: degrees of daily temperature in excess of 34 °C. O3h: ozone high. In parentheses: time lag, in days, in which the association occurs.



**Figure 3.** The Y-axis represents the values of the RRs. The X-axis represents the variables where the statistically significant association is established with the lag in parentheses. Relative risks (RR) with 95% Confidence Interval, corresponding to 10 g/m<sup>3</sup> increase in pollutants and 1 °C increase in T<sub>heat</sub>, of the statistically significant independent variables by cause-specific mortality. In red: values corresponding to days with Saharan dust advection (NAF: North African Dust). In blue: values corresponding to days with no Saharan dust advection (NAF=0). T heat: degrees of daily temperature in excess of 34 °C. O3h: ozone high.



**Figure 4.** The Y-axis represents the values of the RRs. The X-axis represents the variables where the statistically significant association is established with the lag in parentheses. Relative risks (RR) with 95% Confidence Interval, corresponding to 10 g/m<sup>3</sup> increase in pollutants and 1 °C increase in T<sub>heat</sub>, of the statistically significant independent variables by cause-specific hospital admissions. In red: values corresponding to days with Saharan dust advection (NAF: North African Dust). In blue: values corresponding to days with no Saharan dust advection (NAF=0). T<sub>heat</sub>: degrees of daily temperature in excess of 34 °C. O<sub>3h</sub>: ozone high.