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1 SHORT-TERM EFFECTS OF AIR POLLUTION AND NOISE ON EMERGENCY HOSPITAL
2 ADMISSIONS IN MADRID AND ECONOMIC ASSESSMENT

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28 **ABSTRACT**

29 **Introduction:** The aim of this study was to study the effect of air pollution and noise has on the
30 population in Madrid Community (MAR) in the period 2013-2018, and its economic impact.

31 **Methods:** Time series study analysing emergency hospital admissions in the MAR due to all
32 causes (ICD-10: A00-R99), respiratory causes (ICD-10: J00-J99) and circulatory causes (ICD-10:
33 I00-I99) across the period 2013-2018. The main independent variables were mean daily PM_{2.5},
34 PM₁₀, NO₂, 8-hour ozone concentrations, and noise. We controlled for meteorological variables,
35 Public Holidays, seasonality, and the trend and autoregressive nature of the series, and fitted
36 generalised linear models with a Poisson regression link to ascertain the relative risks and
37 attributable risks. In addition, we made an economic assessment of these hospitalisations.

38 **Results:** The following associations were found: NO₂ with admissions due to natural (RR: 1.007,
39 95% CI: 1.004 – 1.011) and respiratory causes (RR: 1.012, 95% CI: 1.005 – 1.019); 8-hour ozone
40 with admissions due to natural (RR: 1.049, 95% CI: 1.014 – 1.046) and circulatory causes (RR:
41 1.088, 95% CI: 1.039 – 1.140); and diurnal noise (L_{Aeq7-23h}) with admissions due to natural (RR:
42 1.001, 95% CI: 1.001 – 1.002), respiratory (RR: 1.002, 95% CI: 1.001 – 1.003) and circulatory
43 causes (RR: 1.003, 95% CI: 1.002 – 1.005). Every year, a total of 8246 (95% CI: 4580 – 11905)
44 natural-cause admissions are attributable to NO₂, with an estimated cost of close on €120 million
45 and 5685 (95% CI: 2533 – 8835) attributed to L_{Aeq7-23h} with an estimated cost of close on €82
46 million.

47 **Conclusions:** Nitrogen dioxide, ozone and noise are the main pollutants to which a large number
48 of hospitalisations in the MAR are attributed, and are thus responsible for a marked
49 deterioration in population health and high related economic impact.

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53 INTRODUCTION

54 Air pollution is a public health problem which affects a large part of the world population
55 (Shaddick et al., 2020). The World Health Organisation (WHO) sets a series of air-pollutant
56 concentration thresholds above which morbidity and mortality increase to an appreciable
57 degree (WHO global air quality guideline, 2021). The effect that pollution has on the respiratory
58 system is well known, producing exacerbations of asthma (Zheng et al., 2021) increased risk of
59 developing chronic obstructive pulmonary disease and lung cancer (Nazar & Niedoszytko, 2022)
60 (Bălă et al., 2021). In the circulatory system it causes coronary syndromes, cerebrovascular
61 diseases and heart failure among others (Joshi et al., 2022) (Qin et al., 2021). There are authors
62 who report that the effects on the human body of air pollution due to NO₂ and PM have no
63 threshold that can be regarded as safe, and that any level of exposure amounts to a health risk
64 (Stafoggia et al., 2022) (Strak et al., 2021). The European Environment Agency's latest report
65 states that 94-99% of the population of the 28 European Union Member States have been
66 exposed to nitrogen dioxide (NO₂), fine particulate matter (PM) and ozone (O₃) levels in excess
67 of the WHO 2019 guideline values (European Environment Agency, 2021).

68 In addition to chemical pollution, the WHO rates acoustic pollution as one of the main
69 environmental risks for health, with a considerable related disease burden in Europe (WHO
70 Regional Office for Europe, 2018). The European Environmental Agency's most recent report
71 points out that one out of every five Europeans is exposed to noise levels that could be seriously
72 harmful to health (European Environment Agency, 2021). Although there are recent research
73 linking noise to depression, risk of cardiovascular disease (Manukyan, 2022) (Münzel et al.,
74 2021) and chronic obstructive pulmonary disease (Liu et al., 2021), the health effects of noise
75 have not been as widely studied as those of chemical pollution.

76 Both chemical and noise pollution tends to affect extremes of age to a greater degree
77 (Gupta et al., 2018) (United Nations Environment Programme, 2022). Affection on genes
78 involved in the inflammatory response has been described if exposure to pollution occurs during

79 pregnancy and early life (Isaevska et al., 2021) and an increased risk of dementia among the
80 elderly (Costa et al., 2020) (Delgado-Saborit et al., 2021).. Hence, a study on the effect of
81 pollution, both chemical and acoustic, by age group, could prove highly useful and would help
82 in designing plans that were more in line with population needs.

83 In recent decades, European countries have implemented policies to regulate air
84 pollution (Rodrigues et al., 2021). This has allowed the average concentration of PM_{2.5} in all
85 European countries to have decreased by 44% by in 2019 compared to 1990 (Juginović et al.,
86 2021). In Spain, an annual limit for PM_{2.5} concentration came into force in 2015, which until then
87 had been an objective and not a limit value (Evaluación de la calidad del aire en España, 2015).

88 Spain's large expanse of territory, the climatic differences between its regions, and the
89 continuous changes in the socio-demographic characteristics of the population can amount to
90 an important limitation in large-scale studies that group data on very extensive regions, by
91 rendering extrapolation of results to specific cities difficult. What is therefore required are
92 studies on limited regions, using updated socio-demographic data. This would allow for specific
93 air-quality improvement plans focused on the most vulnerable population to be implemented
94 in cities, with the aim of achieving an improvement in the quality of life of their inhabitants.

95 From the stance of improvement in the quality of life, rather than measuring
96 attributable or premature mortality, it is more useful to measure morbidity based on hospital
97 admissions, since the latter is a clear indicator of the lack of a healthy lifestyle. Moreover, this
98 would allow for a more accurate quantification of healthcare costs. Despite its usefulness,
99 however, there are few studies that establish the economic impact associated with hospital
100 admissions attributed to air pollution, both chemical and acoustic.

101 The aim of this study was therefore to analyse the effect of chemical and acoustic
102 pollution on hospital admissions of the resident population of the Madrid Autonomous Region
103 (MAR) (*Comunidad Autónoma de Madrid*), by quantifying admissions attributable to pollution
104 and their related economic cost.

105 **METHODS**

106 **Study design**

107 We conducted an ecological, retrospective, longitudinal, time series study. Unscheduled
108 hospital admissions to hospitals in the MAR were linked to short-term air and acoustic pollution,
109 controlling for the effect of different confounding variables, whether meteorological or control
110 variables of seasonality and the trend and autoregressive nature of the series.

111 **Dependent variable**

112 As our dependent variable, we took daily unscheduled admissions as per the International
113 Classification of Diseases, 10th edition (ICD-10), excluding accidents. For analysis purposes, the
114 following were considered: admissions due to all causes (ICD-9:1-799; ICD-10: A00-R99) and,
115 separately, admissions due to respiratory (ICD-9: 460-487; ICD-10: J00-J99) and circulatory
116 causes (ICD-9: 390-459); ICD-10: I00-I99), registered in the MAR across the period 1 January
117 2013-31 December 2018. These data were obtained under a National Statistics Institute
118 (*Instituto Nacional de Estadística/INE*) confidentiality protocol. The data were broken down by
119 age group (<2 years, 2 to 10 years, 11 to 64 years, 65 to 79 years, and 80+ years).

120 **Independent variables**

121 The main independent variables were mean daily air pollution concentrations measured in
122 $\mu\text{g}/\text{m}^3$ and acoustic pollution measured in dB(A). Air quality data were sourced from the Ministry
123 for Ecological Transition and Demographic Challenge (*Ministerio para la Transición Ecológica y*
124 *Reto Demográfico*) for the same period (1 January 2013-31 December 2018). Based on the data
125 set from all the measuring stations in the MAR, we calculated the mean daily levels of particulate
126 matter having a diameter of 2.5 microns or less ($\text{PM}_{2.5}$), particulate matter having a diameter of
127 10 microns or less (PM_{10}), nitrogen dioxide (NO_2), and 8-hour ozone ($\text{O}_{3\text{oct}}$: maximum value
128 registered between 8 and 16 hours).

129 Acoustic pollution data were supplied by the Madrid Municipal Permanent Acoustic
130 Pollution Monitoring Grid (*Red Fija de Control de la Contaminación Acústica*) and the AENA

131 Airports Acoustic Pollution Monitoring Grid. The data from these stations were then used to
132 obtain the mean equivalent diurnal noise ($L_{Aeq7-23h}$), nocturnal noise ($L_{Aeq23-7h}$) and daily noise
133 (L_{Aeq24h}) levels.

134 Due their importance as confounding variables, we used meteorological data drawn
135 from readings taken at the Madrid-Retiro Observatory (MAR reference observatory) for the
136 same period and furnished by the State Meteorological Agency (*Agencia Estatal de*
137 *Meteorología/AEMET*). We recorded the daily values of maximum and minimum temperature
138 ($^{\circ}C$), air pressure (hPa), number of hours of sunlight per day, wind (km/h) and relative humidity
139 (%). Temperature was included as two transformed variables —for low (Carmona et al 2016) and
140 high temperature (Diaz et al 2015)— that converted lineal their non-lineal association with
141 dependant variables. As high temperature is associated with short-term impacts, we included
142 lagged variables of order 4 in this case (Diaz et al 2015). As low temperature is associated with
143 long-term impacts, we include lagged variables of order 14 in their case (Díaz et al 2005). The
144 relationship with relative humidity was linear, so no transformation was necessary. In this case,
145 lagged variables of order 14 were included in the models. The geographical distribution of their
146 stations could be consulted in supplementary materials figure 1.

147 The following control variables were also included: annual (365 days), six-monthly (180
148 days), four-monthly (120 days) and quarterly (90 days) seasonality using the sine and cosine
149 functions. The trend of the series was taken into account using a counter n_1 , which assigns a
150 number to each day of the series, such that n_1 equals 1 for 01/01/2013, n_1 equals 2 for
151 02/01/2013 and so on successively. We created dummy variables for the days of the week and
152 controlled for the autoregressive nature of the dependent variable. In addition, we also
153 controlled for Public Holidays in the MAR across the study period.

154 **Statistical analyses**

155 Initially, we performed a descriptive analysis of all the dependent and independent variables, by
156 calculating the percentage of days on which the WHO's latest 2021 guideline values (WHO global

157 air quality guideline, 2021) and Spanish statutory limit values (RD 102/2011) (*Real Decreto*
158 *102/2011, de 28 de enero, relativo a la mejora de la calidad del aire, 2011*) were exceeded, and
159 controlling for the presence of outliers.

160 According to previous studies, the functional relationships between hospital admissions
161 and the environmental variables analysed display a linear distribution without threshold (Ortiz
162 et al., 2017) (Linares et al., 2018), save in the case of variables of temperature and 8-hour ozone,
163 which show a quadratic distribution (Díaz et al., 1999) (Alberdi et al., 1998). In view of this,
164 parametrisations were only done for these last two variables.

165 In the case of 8-hour ozone, the functional relationship with all-cause hospital
166 admissions was shown to be quadratic. To parametrise the function, the 8-hour ozone level
167 corresponding to the minimum of the quadratic function was calculated by making the pertinent
168 adjustment to a second-order polynomial, with this value being set at $147.5 \mu\text{g}/\text{m}^3$, and a new
169 variable was created to identify values that exceeded the threshold calculated, defined as
170 follows:

$$171 \quad O_{3oct}high = 0, \text{ if } [O_{3oct}] < 147.5 \mu\text{g}/\text{m}^3$$
$$172 \quad O_{3oct}high = O_{3oct} - 147.5 \mu\text{g}/\text{m}^3, \text{ if } [O_{3oct}] > 147.5 \mu\text{g}/\text{m}^3$$

173

174 For temperature, we used the threshold values set by the Ministry of Health for the
175 definitions of heat wave (corresponding to 34°C for the maximum daily temperature) (Ministerio
176 de Sanidad, 2022b) and cold wave in the MAR (corresponding to -2°C for the minimum daily
177 temperature) (Dirección General de Salud Pública, 2017). Two new variables were created,
178 defined as follows:

$$179 \quad Theat = 0, \text{ if } Tmax < 34$$
$$180 \quad Theat = Tmax - 34^\circ\text{C}, \text{ if } Tmax > 34$$

181

$$182 \quad Tcold = 0, \text{ if } Tmin > -2$$

183 $T_{cold} = -2^{\circ}\text{C} - T_{min}, \text{if } T_{min} < -2$

184

185 In the case of noise, given that there are no situations of zero noise, the minimum value
186 of the whole series was established as the threshold, with this being set at 41.7 dB(A) for L_{Aeq7-}
187 $_{23h}$ and 36.3 dB(A) for nocturnal noise.

188 Given that the health effects of independent variables on dependent variables may be
189 lagged in time, we calculated the lags for the variables of air, acoustic and meteorological
190 pollution in order to analyse the delayed effect. In the case of $PM_{2.5}$, PM_{10} , NO_2 and noise, the
191 effect may be lagged by up to 5 days (Ortiz et al., 2017) (Linares et al., 2018), and in the case of
192 ozone by up to 9 days (Díaz et al., 2018). For the difference in pressure, the effect is detected up
193 to 8 days afterwards (González et al., 2001), and the effect of heat may be delayed by up to 5
194 days (Díaz et al., 2002). For relative humidity and cold, there were no previous studies that had
195 analysed these jointly with the remaining meteorological variables, and we therefore allowed
196 for a lag of 14 days.

197 To ascertain the impact of air and acoustic pollution on daily hospital admissions, we
198 fitted generalised linear models (GLMs) with a Poisson regression link and controlled for
199 overdispersion. We applied the backward method, whereby all the variables -both
200 meteorological and pollution-related- are initially included along with their respective time lags.
201 We also included the control variables of seasonality, and trend and autoregressive nature of
202 the series, with those of less statistical significance then being gradually eliminated until all the
203 variables had a significance of $p < 0.05$. Based on the values of the significant estimators, relative
204 risks (RRs) were calculated for every one-unit increase in the independent variables. In the case
205 of ozone, PM_{10} , $PM_{2.5}$ and NO_2 , the increase in risk of hospital admission was calculated for every
206 rise of $10 \mu\text{g}/\text{m}^3$. To estimate the risk of hospital admission attributable to air and acoustic
207 pollution, the above RRs were used to calculate the attributable risks (ARs) as per Coste and
208 Spira's equation (1991) (Coste & Spira, 1991):

209
$$AR = \frac{(RR - 1)}{RR} \times 100$$

210

211 The ARs were then used to calculate the number of admissions attributable to chemical and
212 acoustic pollution in two cases:

213 1. without considering any threshold (in the case of 8-hour ozone, this was considered as
214 from its previously calculated health-impact value).

215 2. based on the guideline values set by the WHO.

216 Lastly, an economic estimate of hospital admissions was drawn up. To this end, the cost of
217 the hospital process was first ascertained by calculating the weighted average of the mean
218 Diagnosis Related Group (DRG) cost, using the APRGRD v.35 grouping system (2019 update).
219 Weightings were assigned in accordance with the relative DRG weights (Ministerio de Sanidad,
220 2022a). We then calculated the number of admissions among the gainfully employed population
221 (INE, 2022a) (INE, 2022b) and the cost of sick leave (Navarro-Espigares JL., 2021), and added the
222 cost of the hospital process to the cost of sick leave.

223 All statistical analyses were performed using the IBM SPSS Statistics v.28 and STATA/BE
224 v17.0 and Microsoft Excel 2019 software packages.

225

226 RESULTS

227 Emergency hospital admissions from 1 January 2013 to 31 December 2018 (total of 2191 days)
228 were analysed. The distribution by type of admission and age group is shown in Table 1. The
229 mean daily number of all-cause admissions (ICD: A00-R99) was 956: of these, 19.4% were
230 admissions due to respiratory causes (ICD-10: J00-J99) and 15.6% were admissions due to
231 circulatory causes (ICD-10: I00-I99). The descriptive analysis of air and acoustic pollution and
232 meteorological conditions is summarised in Table 2.

233 Table 3 shows the percentage of days on which the WHO 2021 (WHO global air quality
234 guideline, 2021) and Spanish statutory thresholds (RD 102/2011) (*Real Decreto 102/2011, de 28*

235 *de enero, relativo a la mejora de la calidad del aire, 2011)* were exceeded. All the pollutants
236 analysed exceeded the WHO threshold at some time. In the case of PM_{2.5}, the threshold (15
237 µg/m³) was exceeded on 11-19% of days in the 2013-2018 series, while for NO₂ the threshold
238 (25 µg/m³) was exceeded on 41-68% of days. In the case of 8-hour ozone, the WHO-
239 recommended threshold (100 µg/m³) was exceeded on 21-29% of days, and the Spanish
240 statutory threshold (120 µg/m³) was exceeded on 4-11% of days.

241 The RRs and ARs, along with their corresponding lags, which proved statistically
242 significant ($p < 0.05$) in the Poisson regression models for hospital admissions due to natural,
243 respiratory and circulatory causes, are shown in Figure 1. Whereas NO₂ had a significant effect
244 on natural and respiratory-cause admissions, 8-hour ozone was linked to admissions due to
245 natural and circulatory causes. No statistically significant association was however found for PM
246 for any type of admission. With regard to noise, a significant association was observed for all
247 three types of admissions, with the most relevant being that detected for circulatory-cause
248 admissions attributable to L_{Aeq7-23h}.

249 Figure 2 shows the significant independent variables with the models broken down by
250 age group for natural-cause admissions. The pollutant with the greatest effect on admissions
251 was 8-hour ozone in the under-2 age group.

252 The models constructed in our study made it possible to quantify admissions
253 attributable to chemical and acoustic pollution, along with their economic assessment, which
254 can be seen in Tables 4 and 5. The results show that every year, 8246 (95% CI: 4580 – 11905)
255 natural-cause admissions are attributable to overall exposure to NO₂, which accounts for 2.4%
256 of all hospitalisations. These admissions amount to an annual hospital expenditure of close on
257 120 million euros. Moreover, if the thresholds established by the WHO had been complied with,
258 2388 (95% CI: 1326 – 3447) annual natural-cause admissions attributable to exposure to NO₂
259 could have been prevented. In the case of respiratory causes, 2728 (95% CI: 1231 – 4219) annual
260 admissions were attributed to NO₂, accounting for 4% of all hospitalisations. Applying the WHO

261 guideline values, out of this total, 862 (95% CI: 389 – 1334) annual admissions could have been
262 prevented.

263 The pollutant with the second highest number of attributable admissions was noise.
264 Every year, 5685 (95% CI: 2533 – 8835) natural-cause admissions are attributable to exposure
265 to acoustic pollution on the same day, accounting for 1.6% of all hospital admissions. These
266 admissions amount to more than 82 million euros in annual hospital expenditure. The effect of
267 noise is even greater in respiratory- and circulatory-cause admissions, 2.9% and 4.47% of which
268 are respectively attributed to acoustic exposure on the same day.

269

270 **DISCUSSION**

271 The effect of air pollution on mortality has been widely studied in recent years (Stafoggia et al.,
272 2022) (Orellano et al., 2020), but to be able assess the effect exerted by pollution on health it is
273 necessary to analyse morbidity as reflected in emergency hospital admissions. In our work, NO₂
274 was found to be associated with an elevated risk of natural and respiratory cause admissions,
275 and 8-hour ozone was found to be associated with an elevated risk of natural and circulatory
276 cause admissions. These findings are closely related to both pollutants exceeding WHO
277 recommended thresholds 20-70% of the days. A study conducted in Beijing corroborates the
278 fact that NO₂ and ozone are the pollutants with greatest impact on health (Zhong et al., 2019).

279 Yet, other studies describe different magnitudes in the associations between hospital
280 admissions and pollution. A study of a French cohort followed up for 12 years, with mean PM_{2.5}
281 concentration levels of 10.6 µg/m³, NO₂ concentration levels of 25.8 µg/m³, and ozone
282 concentration levels of 94.4 µg/m³ -all similar to those described in our study- obtained the
283 following RRs for natural-cause admissions: 1.107 (95% CI: 1.079 – 1.136) for PM_{2.5}; 1.029 (95%
284 CI: 1.002 – 1.057) for NO₂; and 1.008 (95% CI: 0.974 – 1.044) for ozone (Sanyal et al., 2018).
285 Similar results were found in a recent meta-analysis which studied countries around the world
286 and reported RRs for respiratory-cause admissions of 1.014 (95% CI: 1.008 – 1.020) for NO₂,

287 1.008 (95% CI: 1.005 – 1.011) for 8-hour ozone, and 1.010 (95% CI: 1.001 – 1.020) for SO₂ (Zheng
288 et al., 2021). Another recent meta-analysis with data on over 23 million participants described
289 that exposure to PM_{2.5}, PM₁₀, SO₂, NO₂, CO and ozone was associated with a statistically
290 significant increase in risk of hospital admission due to cerebrovascular disease for all pollutants
291 (Niu et al., 2021). Contrary to what has been shown in these studies, in our work the RR of
292 natural causes due to 8-hour ozone is much higher than the RR due to NO₂, however, admissions
293 attributable to NO₂ are approximately 100 times higher than admissions attributed to 8-hour
294 ozone. This is because in Madrid the functional relationship of 8-hour ozone with hospital
295 admissions is quadratic. The minimum value of atmospheric ozone corresponding to the
296 minimum value of admissions is 147.5 µg/m³, a very high value that only occurs in 0.3% of the
297 days of the series studied.

298 When it comes to the percentage of admissions attributable to pollution, there are also
299 differences in the studies published to date. A systematic review of 43 papers addressing the
300 effects of NO₂ on hospital admissions describes a 0.57% (95% CI: 0.33 – 0.82) increase in
301 respiratory-cause admissions for every daily rise of 10 µg/m³ in NO₂ (Mills et al., 2015). In
302 contrast, a recent study conducted in Casablanca reported that an a 10 µg/m³ rise in NO₂
303 concentrations increased circulatory-cause admissions by 9% at lag 3, and respiratory-cause
304 admissions by 3% on the same day (Nejjari et al., 2021). Seen against our Madrid-based study,
305 this latter value is the most similar, with a 2% increase in respiratory admissions for NO₂ on the
306 same day, though with no relationship being found between NO₂ and circulatory-cause
307 admissions. All these differences may be due to a lack of homogeneity between the populations
308 compared, since many health determinants, such as socio-economic level, race, education, type
309 of dwelling, etc., were not taken into account in most of the studies.

310 Furthermore, when analysing the effect of pollution on the human body, one has to take
311 into account other factors, such as temperature and other meteorological variables, which are
312 intercorrelated (Ahad et al., 2020). For instance, the presence of NO₂, solar radiation and high

313 temperatures are required for the formation of ozone (WHO global air quality guideline, 2021).
314 As regards temperatures, the effect *per se* of heat and cold waves on health is well documented
315 (Wondmagegn et al., 2021) (López-Bueno et al., 2020), though recent studies undertaken in
316 Spain show that mortality attributable to heat is gradually decreasing, due to adaptation to
317 temperatures (López-Bueno et al., 2021) (Follos et al., 2020). For this reason, and in order to
318 study the effect of atmospheric pollution as reliably and accurately as possible, meteorological
319 variables have been included in our research, including heat and cold waves. This means that
320 studies that do not take these variables into account may offer results that differ from ours.

321 Our study observed that, at present, PM are not statistically significantly associated with
322 any type of short-term emergency hospital admission in the MAR. This may be due to the fact
323 that the mean annual concentrations of PM_{2.5} and PM₁₀ in Madrid fell from 13 and 25 µg/m³ in
324 2009 to 9.9 and 17.5 µg/m³ in 2019 respectively (Dirección General de Medioambiente
325 Sostenibilidad, 2019). One of the possible reasons for this drop is that in 2015 the annual limit
326 for this pollutant entered into force (Evaluación de la calidad del aire en España, 2015). In
327 Madrid, in the period 2013-2018, the SO₂ and CO pollution levels were very much below the
328 values regarded as a health risk, with average annual concentrations of 2 µg/m³ and 0.4 µg/m³
329 respectively (Dirección General de Medioambiente Sostenibilidad, 2019), which is in fact why
330 these pollutants were not included in our analyses.

331 The effect of pollution varies according to the age of the population, with children and
332 the elderly being most vulnerable, so that an analysis broken down by age group yields valuable
333 information. Our study found that the age groups with the highest risk of hospital admission due
334 to pollution were children under 2 years for 8-hour ozone, followed by the elderly aged from 65
335 to 79 years for the same pollutant. As mentioned above, it should be borne in mind that there
336 were only 7 days in the entire series with elevated ozone concentrations resulting in very high
337 RRs. Moreover, the negligible number of hospitalisations occurring in the under-2 age group
338 means that the RRs obtained are higher than in other age groups. Even so, a Texas study also

339 described the important effect that pollution has on children, reporting that at 4 days of 8-hour
340 ozone exposure, children aged 5 to 14 years had a higher risk of hospitalisation due to asthma
341 than did the older population, with an RR of 1.047 (95% CI: 1.025 – 1.069) (Zu et al., 2017).

342 The effect of NO₂ concentrations is seen in all age groups, with a weaker association in
343 terms of RR but with a greater impact on hospital admissions overall, this being the chemical
344 pollutant to which most hospital admissions are attributed. One study undertaken in China
345 found a statistically significant association for NO₂ at lag 5, and hospital admissions due to
346 pneumonia in children under 15 years of age, with the effect being most marked in infants under
347 1 year of age (Li et al., 2018).

348 In addition to chemical-type air pollution, acoustic pollution has an equally relevant role
349 in health. Its effect on short-term emergency hospital admissions is shown by the fact that our
350 study found a statistically significant association between L_{Aeq7-23h} and all types of admissions,
351 and for all age ranges, except children under the age of 2, of whom there were fewer.
352 Furthermore, noise is the pollutant with which the highest proportion of attributable admissions
353 is related, even higher than that recorded for NO₂, ozone or PM. These findings are in line with
354 those reported by other authors, such as Halonen et al., who report that L_{Aeq7-23h} increases the
355 risk of hospitalisation due to cardiovascular disease in adults, and that nocturnal noise is
356 responsible for the same effect in the elderly (Halonen et al., 2015). As mentioned above,
357 children and the elderly are more vulnerable to the effect of the noise than is the rest of the
358 population. A previous study in Madrid conducted on children under the age of 10 years
359 described RRs of 1.02 (95% CI: 1.01 – 1.04) for natural-cause admissions and 1.05 (95% CI: 1.01
360 – 1.08) for respiratory- cause admissions (Linares et al., 2006) consistent with what was found
361 in our study. More recent studies conducted in Madrid have respectively reported associations
362 between exposure to noise and admissions due to suicide and anxiety (Díaz et al., 2020), and
363 admissions due to exacerbations of symptoms of dementia (Linares et al., 2017).

364 Mention should also be made of other studies undertaken in Madrid, in which traffic
365 noise has been linked to morbidity and mortality due to circulatory and respiratory causes
366 (Tobías et al., 2015) (Recio et al., 2017). The common source for a very high proportion of NO₂
367 and noise pollution is road traffic, thereby making vehicles the main cause of hospital admissions
368 due to circulatory and respiratory causes in Madrid (Navares et al., 2020). This highlights the fact
369 that, while noise is a pollutant with health effects similar to those of the other chemical
370 pollutants, scant research has targeted this pollutant, thus calling for more studies to continue
371 assessing the effect of noise on health.

372 In addition to the marked impact which pollution has on population health, attention
373 should be also drawn to the high related economic cost. Our study shows that NO₂ and noise
374 are responsible for the highest number of attributable admissions, accompanied by an
375 enormous hospital cost. From 2004 to 2006, the Aphekom project studied the impact of
376 pollution in 25 European cities, and estimated that, if the threshold values annually
377 recommended by the WHO in Europe had been observed, health systems would have saved 31
378 billion euros (Pascal et al., 2013). It should be borne in mind here that these data are twenty
379 years old, so the current cost could be expected to be considerably higher. The lack of recent
380 studies that calculate the healthcare cost associated with air and acoustic pollution highlights
381 the need for further research to improve the welfare of the population.

382 **Limitations**

383 This study has a number of limitations. While the methodology used is standard for these types
384 of studies (Samet et al., 2000), it can nevertheless give rise to a number of biases. To minimise
385 these, we controlled for a range of variables, such as seasonality and the trend and
386 autoregressive nature of the series. As this is an ecological study, individual inferences cannot
387 be drawn from the results obtained, due to the ecological fallacy (Piantadosi et al., 1988). The
388 study design only allows for collection of pollution data from a given number of stations, thus
389 not representing real exposure at an individual level but instead obtaining average values of

390 daily concentrations. This approach was adopted to cover the possibility of residents who go to
391 work at different places and are thus exposed to different levels of exposure. In the case of
392 meteorological variables, data were obtained from a single observatory, but it should
393 nevertheless be noted that it is the designated reference observatory for the MAR (Díaz et al.,
394 2002).

395 Madrid is an enough big city as to consider suburban areas or zones of influence
396 associated with different stations. The exposure levels used were based on exposures
397 determined on the basis of readings taken by external monitors and then averaged, with the
398 result that they are not measures which represent individual exposure. Another bias from the
399 monitors used can be the heterogeneity in the type, they are mostly urban but occasionally we
400 used background type. Even so, this is a commonly used methodology in these types of studies
401 (Samet et al., 2000). However, much of this residual confusion is controlled by inclusion in the
402 model of variables such as: trend of the series, day of the week, annual, six-monthly and
403 quarterly seasonalities and the autoregressive nature of the series. No specific validation was
404 carried out to assess the representativeness of spatial variability in air pollutants: our study
405 suffered from Berkson type measurement error, among other biases associated with an
406 ecological exposure, as is common in most time-series studies of air pollution, which leads to no
407 or little bias but decreases statistical power. At all events, most air-pollution studies address
408 the misalignment problem (albeit only implicitly), by using a two-stage modelling procedure, or
409 plug-in approach, where predictions from an exposure model (first stage) are used as covariates
410 in a health-effect model (second stage) (Barceló et al., 2016).

411 Furthermore, there are significant correlations between particulate matter and
412 between particulate matter and the high ozone variable (supplementary table). This problem
413 has been controlled in the modelling process by including the control variables and excluding
414 non-significant meteorological variables from the model.

415 A further limitation of this study is that use was made of hospital admission data which
416 may have undergone coding errors: the foreseeable effect is minimal, however, since data
417 sourced from official sources (National Statistics Institute/INE) were used across the entire study
418 period.

419 **Strengths**

420 To our knowledge, this is one of the most up-to-date studies conducted in Spain in terms of the
421 different types of pollutants analysed and their impact on short-term emergency hospital
422 admissions. Moreover, many atmospheric variables and possible confounding factors were
423 controlled for, in order to minimise the possibility of biases.

424 The study included an analysis of acoustic pollution, an aspect that is still little studied,
425 thereby contributing to the generation of impact data on this environmental variable. No recent
426 papers were found on the effect had by traffic-related acoustic pollution on hospital admissions
427 due to natural causes or major disease groups, despite noise being a risk factor, similar to air
428 pollution, that gives rise to a significant morbidity burden and related economic cost.

429 This study is, moreover, of great interest for national plans that seek to address the
430 environment and health and the development of measurable and quantifiable indicators, since
431 the effects of pollution were analysed at a regional level and their calculation can be
432 extrapolated to other geographical areas of interest.

433 **Conclusion**

434 Despite the fact that air-quality control regulations have led to an improvement in the
435 atmospheric situation in recent years, air pollutant levels continue to be high. The related
436 morbidity is manifested in an excessive number of attributable hospital admissions, with the
437 ensuing loss of the population's quality of life and the great financial cost that this entails. The
438 short-term effect on hospital admissions due to NO₂ and ozone concentrations and noise levels
439 continues to be very high in cities such as Madrid, rendering it necessary to update statutory

440 threshold limits to levels closer to those recommended by the WHO, and to implement new
441 public health policies that would continue to protect the health of the entire population.

442 **Conflicts of interest**

443 The authors declare that there are no conflicts of interest.

444 **Disclaimer**

445 The researchers declare that they have no conflicts of interest that would compromise the
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452

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683

TABLE 1: Description of daily hospital admissions for the 2013-2018 series, Madrid.

Variable	Mean admissions per year	Mean	SD	Minimum	Maximum
All-cause admissions:					
Total	349099	956	169	537	1395
Women	196530	538	89	319	785
Men	152569	418	83	201	650
<2 years	20574	56	20	12	140
2 to 10 years	12481	34	11	6	68
11 to 64 years	152655	418	67	221	573
65 to 79 years	72373	198	45	83	355
80+ years	91016	249	58	103	502
Respiratory-cause admissions:					
Total	67825	186	77	48	554
Circulatory-cause admissions:					
Total	54413	149	34	54	230

SD: standard deviation.

TABLE 2: Description of air and acoustic pollution and the meteorological conditions for the 2013-2018 series, Madrid.

Variable	Mean	SD	Minimum	Maximum
Air pollution				
PM _{2,5} (µg/m ³)	10,3	4,7	3,2	33,1
PM ₁₀ (µg/m ³)	19,0	9,7	2,9	85,7
NO ₂ (µg/m ³)	30,7	14,5	5,8	90,9
O _{3 oct} (µg/m ³)	78,6	28,8	9,9	171,5
Acoustic pollution:				
Diurnal noise (dB(A))	55,8	2,5	41,7	62,1
Nocturnal noise (dB(A))	48	2,6	36,3	61,4
24h noise (dB(A))	54,1	2,4	40,2	60,0
Meteorological conditions:				
Relative humidity (%)	59,7	16,3	19,0	95,2
Atmospheric pressure (hPa)	940,7	6,0	911,8	962,6
T _{max} (°C)	21,1	9,1	2,8	40,0
T _{min} (°C)	11,1	6,8	-3,0	25,9
Wind (km/h)	6,4	3,0	0,0	18,7
No. sunlight hours per day	8,1	4,3	0,0	14,4

SD: standard deviation.

TABLE 3: Percentage of days on which the WHO 2021 and Spanish statutory (RD 102/2011) thresholds were exceeded.

	PM _{2.5}	PM ₁₀	NO ₂	O _{3 oct}
2013				
WHO 2021 threshold	12.6%	0.5%	52.3%	21.1%
Spanish statutory threshold	-	0%	-	4.4%
2014				
WHO 2021 threshold	11.2%	1.9%	55.6%	21.4%
Spanish statutory threshold	-	0.8%	-	5.2%
2015				
WHO 2021 threshold	18.6%	3.3%	64.1%	26.6%
Spanish statutory threshold	-	1.6%	-	10.7%
2016				
WHO 2021 threshold	15.8%	3.3%	65.6%	24.6%
Spanish statutory threshold	-	1.4%	-	6.3%
2017				
WHO 2021 threshold	17.0%	2.2%	68.2%	27.9%
Spanish statutory threshold	-	0.8%	-	8.2%
2018				
WHO 2021 threshold	14.8%	2.2%	40.5%	29.0%
Spanish statutory threshold	-	0.8%	-	8.5%

PM_{2.5}: WHO threshold of 15 µg/m³.

PM₁₀: WHO threshold of 45 µg/m³, Spanish statutory threshold: 50 µg/m³.

NO₂: threshold WHO de 25 µg/m³.

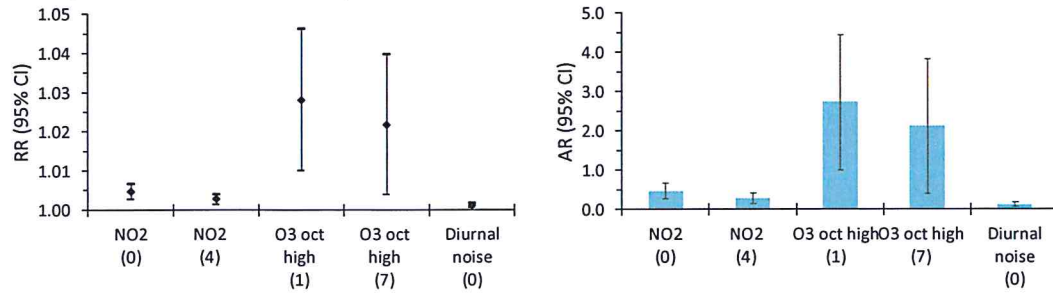
O_{3 oct}: WHO threshold of 100 µg/m³, Spanish statutory threshold: 120 µg/m³.

TABLE 4: Admissions per year attributable to pollutant.

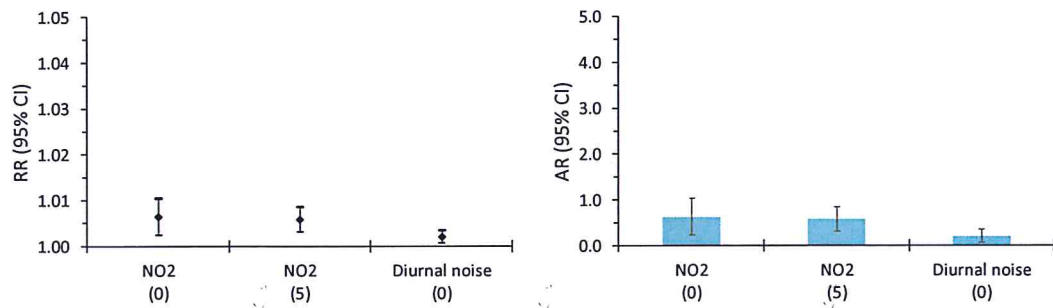
	Mean annual attributable admissions (95% CI)	Mean annual admissions on WHO threshold being exceeded (95% CI)	% attributable admissions
Natural-cause admissions:			
NO ₂ (lag 0)	5191 (2966 - 7412)	1503 (859 - 2146)	1.49%
NO ₂ (lag 4)	3055 (1614 - 4493)	885 (467 - 1301)	0.88%
O _{3 oct high} (lag 1)	31 (11 - 51)	31 (11 - 51)	0.01%
O _{3 oct high} (lag 7)	24 (4 - 44)	24 (4 - 44)	0.01%
Diurnal noise (lag 0)	5685 (2533 - 8835)	-	1.63%
Respiratory-cause admissions:			
NO ₂ (lag 0)	1427 (529 - 2321)	451 (167 - 734)	2.10%
NO ₂ (lag 5)	1301 (702 - 1898)	411 (222 - 600)	1.92%
Diurnal noise (lag 0)	1987 (594 - 3378)	-	2.93%
Circulatory-cause admissions:			
O _{3 oct high} (lag 4)	13 (6 - 19)	13 (6 - 19)	0.02%
Diurnal noise (lag 0)	2432 (1257 - 3605)	-	4.47%

Mean annual admissions on WHO threshold being exceeded, was obtained by calculating the mean number of daily admissions that occurred on days on which the daily threshold value recommended by the WHO was exceeded (O₃: 100 µg/m³, NO₂: 25 µg/m³).

Effect of pollutants on natural-cause hospital admissions:



Effect of pollutants on respiratory-cause hospital admissions:



Effect of pollutants on circulatory-cause hospital admissions:

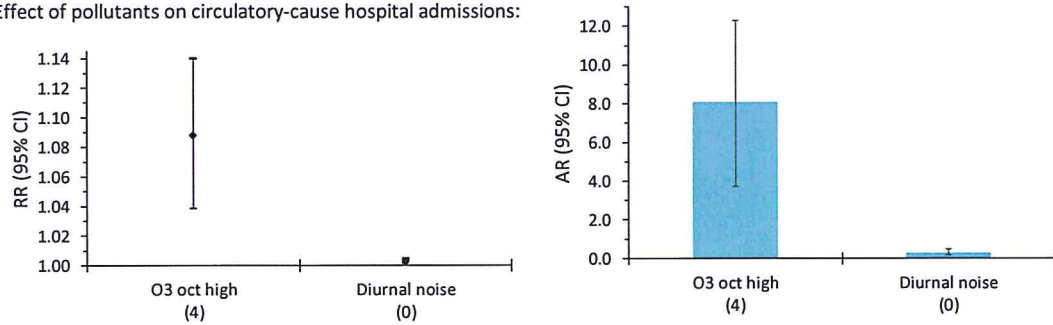


FIG 1. Relative risks (RRs) and attributable risks (ARs) with their respective 95% CIs of the significant independent variables according to type of hospital admission. Increases for every $10 \mu\text{g}/\text{m}^3$ for air pollutants. In the case of ozone, increases for every $10 \mu\text{g}/\text{m}^3$ above the 8-hour ozone threshold of $147.5 \mu\text{g}/\text{m}^3$. Lags shown in brackets.

Effect of pollutants on natural-cause hospital admissions:

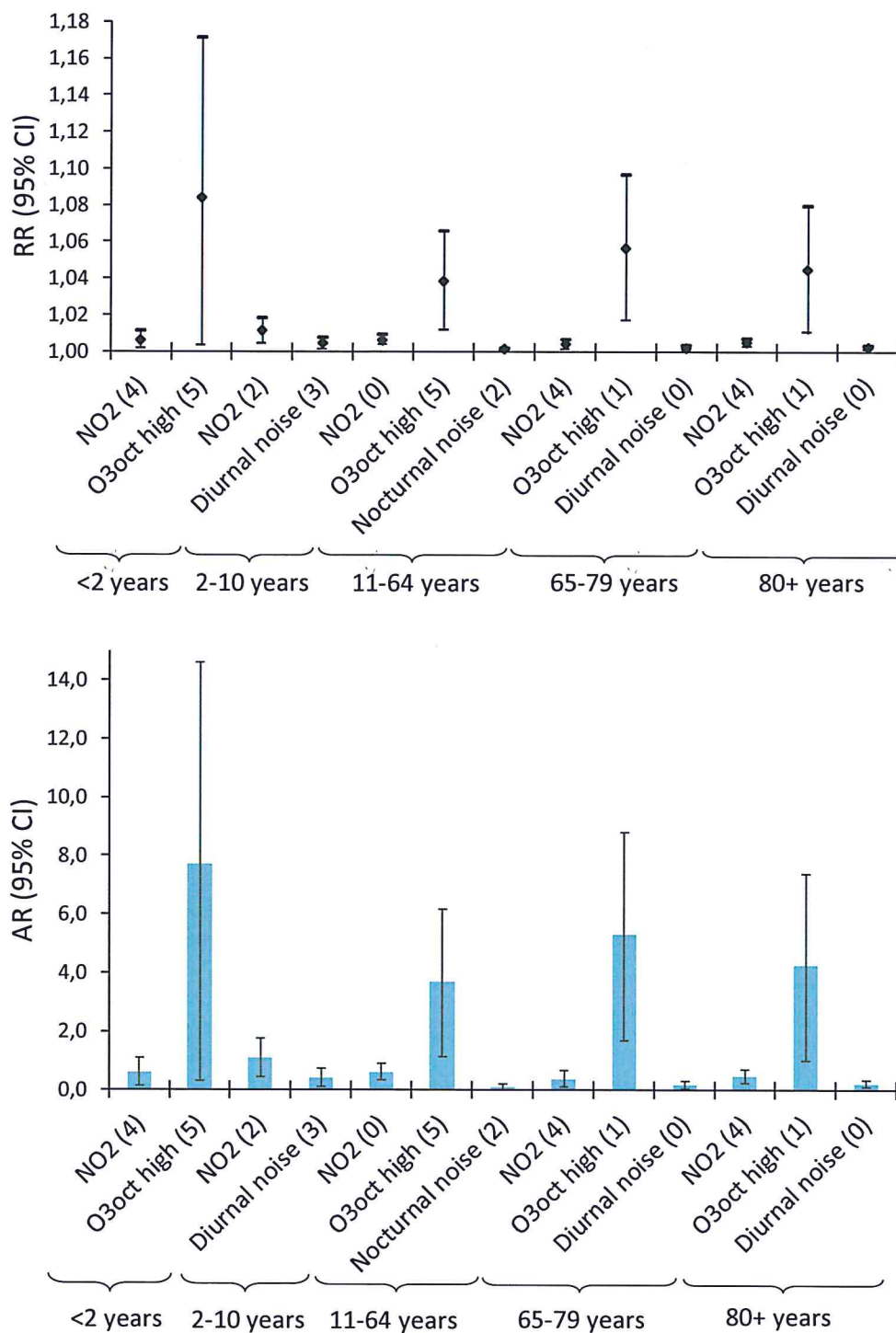


FIG 2. Relative risks (RRs) and attributable risks (ARs) with their respective 95% CIs of the significant independent variables in natural-cause admissions. Increases for every $10 \mu\text{g}/\text{m}^3$ for air pollutants. In the case of ozone, increases for every $10 \mu\text{g}/\text{m}^3$ above the 8-hour ozone threshold of $147.5 \mu\text{g}/\text{m}^3$. Lags shown in brackets.