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Short-term air pollution exposure and hospital admissions for cardiorespiratory diseases in Brazil: A nationwide time-series study between 2008 and 2018

Weeberb J. Requia, Ana Maria Vicedo-Cabrera, Heresh Amini, Gladston Luiz da Silva, Joel D. Schwartz, Petros Koutrakis

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## **Credit Author Statement**

**Weeberb J. Requia:** Conceptualization, Methodology, Software, Data curation, Writing- Original draft preparation, Funding acquisition, Resources.

**Ana Maria Vicedo-Cabrera:** Conceptualization, Methodology, Software, and Writing- Reviewing and Editing

**Heresh Amini:** Methodology, Software, and Writing- Reviewing and Editing

**Gladston Luiz da Silva:** Writing- Reviewing and Editing

**Joel D. Schwartz:** Writing- Reviewing and Editing

**Petros Koutrakis:** Writing- Reviewing and Editing

**Short-term air pollution exposure and hospital admissions for cardiorespiratory diseases in Brazil: a nationwide time-series study between 2008-2018**

**Weeberb J. Requia**

(Corresponding Author)

weeberb.requia@fgv.br

School of Public Policy and Government, Fundação Getúlio Vargas

Brasília, Distrito Federal, Brazil

**Ana Maria Vicedo-Cabrera**

1. Institute of Social and Preventive Medicine, University of Bern, Bern, Switzerland

2. Oeschger Center for Climate Change Research, University of Bern

Bern, Switzerland

**Heresh Amini**

Department of Public Health, University of Copenhagen

Copenhagen, Denmark

**Gladston Luiz da Silva**

Department of Statistics, University of Brasilia

Brasilia, Brazil

**Joel D. Schwartz**

Department of Environmental Health, Harvard TH Chan School of Public Health

Boston, Massachusetts, United States

**Petros Koutrakis**

Department of Environmental Health, Harvard TH Chan School of Public Health

Boston, Massachusetts, United States

1 **Short-term air pollution exposure and hospital admissions for cardiorespiratory diseases in**  
2 **Brazil: a nationwide time-series study between 2008-2018**

3  
4 **Abstract**

5 The established evidence associating air pollution with health is limited to populations from  
6 specific regions. Further large-scale studies in several regions worldwide are needed to support the  
7 literature to date and encourage national governments to act. Brazil is an example of these regions  
8 where little research has been performed on a large scale. To address this gap, we conducted a  
9 study looking at the relationship between daily PM<sub>2.5</sub>, NO<sub>2</sub>, and O<sub>3</sub>, and hospital admissions for  
10 circulatory and respiratory diseases across Brazil between 2008 and 2018. A time-series analytic  
11 approach was applied with a distributed lag modeling framework. We used a generalized  
12 conditional quasi-Poisson regression model to estimate relative risks (RRs) of the association of  
13 each air pollutant with the hospitalization for circulatory and respiratory diseases by sex, age  
14 group, and Brazilian regions. Our study population includes 23,791,093 hospital admissions for  
15 cardiorespiratory diseases in Brazil between 2008 and 2018. Among those, 53.1% are respiratory  
16 diseases, and 46.9% are circulatory diseases. Our findings suggest significant associations of  
17 ambient air pollution (PM<sub>2.5</sub>, NO<sub>2</sub>, and O<sub>3</sub>) with respiratory and circulatory hospital admissions in  
18 Brazil. The national meta-analysis for the whole population showed that for every increase of  
19 PM<sub>2.5</sub> by 10µg/m<sup>3</sup>, there is a 3.28% (95%CI: 2.61; 3.94) increase in the risk of hospital admission  
20 for respiratory diseases. For O<sub>3</sub>, we found positive associations only for some sub-group analyses  
21 by age and sex. For NO<sub>2</sub>, our findings suggest that a 10ppb increase in this pollutant, there was a  
22 35.26% (95%CI: 24.07; 46.44) increase in the risk of hospital admission for respiratory diseases.  
23 This study may better support policymakers to improve the air quality and public health in Brazil.

24  
25 **Keywords:** Air pollution, Hospital admissions, Respiratory diseases, Circulatory diseases, Brazil.

## 26 1. INTRODUCTION

27 Air pollution is a major environmental risk to global health (Cohen et al., 2017). It is well  
28 established that exposure to ambient air pollution is associated with increased hospital admissions  
29 (Carugno et al., 2016; Danesh Yazdi et al., 2021; Phung et al., 2016) and mortality (Burnett et al., 2018;  
30 Schwartz et al., 2021b; Shi et al., 2021). Meta-analysis studies have reported consistent positive  
31 associations worldwide between air pollutants and several causes of hospitalization, including  
32 hospital admissions for respiratory and circulatory diseases (Moore et al., 2016; Requia et al., 2017).  
33 According to the World Health Organization (WHO), about 4.2 million deaths worldwide all-cause  
34 and all-age occur every year due to exposure to ambient air pollution (WHO, 2022). Particles with  
35 a mass aerodynamic diameter below 2.5  $\mu\text{m}$  (PM<sub>2.5</sub>), itself contributes to approximately two  
36 million premature deaths per year, ranking it as the 13<sup>th</sup> leading cause of worldwide mortality  
37 (Lozano et al., 2012). The WHO estimates that 80% of outdoor air pollution-related premature deaths  
38 are associated with ischemic heart disease and strokes; and 14% with chronic obstructive  
39 pulmonary disease and acute lower respiratory infections (WHO, 2022). More recently, numerous  
40 investigations have indicated consistency in the link between health outcomes and other air  
41 pollutants, including gaseous pollutants such as nitrogen dioxide (NO<sub>2</sub>) and tropospheric ozone  
42 (O<sub>3</sub>) (Danesh Yazdi et al., 2021; Schwartz et al., 2021b, 2021a; Shi et al., 2021; Wei et al., 2021; Yazdi et  
43 al., 2021).

44 Overall, this established evidence associating air pollution with health is limited to  
45 populations from specific regions. A systematic review by (Requia et al., 2017) found that most of  
46 the evidence is based on populations in the United States, some countries in Europe, and China.  
47 Further large-scale studies in several other populated countries are needed to support the literature  
48 to date. Numerous regions worldwide face critical challenges related to poor air quality and public  
49 health, and governments are reluctant to embark on expensive controls based on studies from other  
50 nations. Epidemiological studies are essential in these regions, which could provide evidence for  
51 policymakers with the objective of improving the air quality and public health.

52 Brazil is an example of these regions with critical challenges and where little research has  
53 been performed at a large scale (national-level spatial resolution). We searched in PubMed and  
54 Web of Science using the following keywords: “air pollution”, “health effects”, “human  
55 exposure”, and “Brazil”. We found only 3 studies at national level in Brazil looking only at the  
56 effects from PM<sub>2.5</sub>, in which one investigation focused on the association between PM<sub>2.5</sub> and

57 cancer hospitalizations (Yu et al., 2021), another one focused on the link between PM<sub>2.5</sub> and cancer  
58 mortality (Yu et al., 2022), and the third one looked at the association between hospital admissions  
59 and wildfire-related PM<sub>2.5</sub> (Ye et al., 2021).

60       Among the critical challenges in Brazil related to air quality and public health, we highlight  
61 i) about 25% of the Brazilian schools are located within a distance  $\leq 250\text{m}$  of a major roadway,  
62 have  $\geq 2\text{km}$  of roadway within a buffer of 1km, and have  $\geq 7$  wildfires records within a buffer of  
63 10km (Requia et al., 2021); ii) Brazil is a very fire-prone region where, according to the National  
64 Institute of Spatial Research - INPE (<http://queimadas.dgi.inpe.br/queimadas/>), between  
65 January/2020 and August/2020 there were about 120,000 km<sup>2</sup> of burned area. Wildfires emit  
66 substantial amounts of air pollutants that can travel over large distances, affecting air quality and  
67 human health far from the originating fires (Youssef et al., 2014). On an equal-mass basis, wildfire-  
68 related PM<sub>2.5</sub> may be more toxic than ambient PM<sub>2.5</sub> in the same region during non-fire periods,  
69 due to the formation of secondary pollutants as a result of the atmospheric photochemistry. Fine  
70 particulate matter (PM<sub>2.5</sub>) is the major pollutant emitted by wildfires. About 12-16% of global  
71 wildfire-related particulate emissions occur across Brazil (Reddington et al., 2015); iii) Third, Brazil  
72 has a lack of control and monitoring policies leading to increased air pollution (Barcellos et al.,  
73 2009; Huneus et al., 2020; Renata et al., 2015); and iv) Brazil also has a considerable difference  
74 in the quality of health/environment and healthcare across different populations (influencing  
75 health/environment equity in negative ways), which is a critical determinant of the health impacts  
76 of air pollution (Rappold et al. 2012).

77       Therefore, more research is needed to explore the nationwide effect of air pollution on  
78 health in Brazil, and to investigate the health outcomes associated with multiple air pollutants in a  
79 region with challenges related to air pollution sources, environmental characteristics, and public  
80 health conditions. To address this gap, we have conducted a study looking at the relationship  
81 between daily PM<sub>2.5</sub>, NO<sub>2</sub>, and O<sub>3</sub>, and hospital admissions for circulatory and respiratory  
82 outcomes across Brazil between 2008 and 2018.

83

## 84 **2. METHODS**

### 85 **2.1. Study population and outcome assessment**

86       The hospital admission data were provided by the Ministry of Health in Brazil. The data  
87 encompass individual hospital admissions records in Brazil between 1<sup>st</sup> January 2008 and 31<sup>st</sup>

88 December 2018. During the study period, there were a total of 129,978,694 records of hospital  
89 admissions (all cases) in Brazil.

90 Hospital admission information included event date, home municipality, age, sex, and  
91 principal diagnosis according to the International Classification of Diseases, version 10 (ICD-10)  
92 codes. In this study, we selected only the hospitalizations for respiratory (ICD-10 codes J00-J99)  
93 and cardiovascular (ICD-10 codes I00-I99) diseases. This filter resulted in 12,621,970 (9.71%)  
94 records of hospital admissions for respiratory outcomes and 11,169,123 (8.59%) hospital  
95 admissions for circulatory diseases.

96 We grouped the data by summing the number of hospital admissions for every municipality  
97 and date in the study period. The hospital admission counts were also divided by sex, age group  
98 (15-25, 26-35, 36-45, 46-55, 56-65, >65 years old), health outcome (respiratory, circulatory, and  
99 cardiorespiratory), and region. Each group representing the counts of hospital admissions is  
100 composed by a complete time series of equality-spaced observations taken in each municipality  
101 and each day in the period 2018-2018. We observed some inconsistencies in the data (missing  
102 values for several municipalities) for the age group <15 years old. Therefore, this age group was  
103 not included in the analyses. There are 5,572 municipalities in Brazil, representing the smallest  
104 areas considered by the Brazilian political system. The municipalities are grouped into five  
105 regions: North, Northeast, Midwest, Southeast, and South. In Supplementary Materials (Figure  
106 S1), we show the spatial distribution of all municipalities and regions in Brazil.

107

## 108 **2.2. Exposure assessment**

109 Air pollution data were obtained from ensemble models that used various satellite  
110 observations. We studied daily PM<sub>2.5</sub> ( $\mu\text{g m}^{-3}$ ), NO<sub>2</sub> (ppb), and O<sub>3</sub> (ppb) concentrations from 2008  
111 to 2018. The data was accessed from the European Centre for Medium-Range Weather Forecasts  
112 (ECMWF). The ECMWF operates services related to meteorology and air pollution covariates and  
113 implements the Copernicus Atmosphere Monitoring Service (CAMS) on behalf of the European  
114 Union, including CAMS-Reanalysis and CAMS Near Real Time (CAMS-NRT) forecasts. The  
115 CAMS service runs ensemble models using several satellite observations and emission inventories,  
116 amongst other predictors. We obtained the data at a spatial resolution of 0.125 degrees  
117 (approximately 12.5 km) and a temporal resolution of 6 hours, including daily estimates for 00,  
118 06, 12, and 18 UTC (Universal Time Coordinated). In our analyses, we used CAMS-Reanalysis

119 for the period between 2008- 2017 and CAMS-NRT for the year 2018. We calculated the daily  
120 mean temporal resolution for each pollutant. Finally, we aggregated the air pollution data by the  
121 municipality, considering the geographical location of the headquarters of each municipality in  
122 Brazil.

123

### 124 **2.3. Confounders**

125 Meteorological data were treated as potential confounders. The data was retrieved from  
126 reanalysis data that used satellite remote sensing and was accessed from the ECMWF. Weather  
127 data included surface temperature (°C), humidity (%), and precipitation (mm/day). Temperature  
128 and humidity were derived from Era-Interim reanalysis, with a spatial resolution of 0.125 degrees  
129 and temporal resolution of 6 hours. This reanalysis was performed by the ECMWF. Precipitation  
130 data was accessed from the Climate Prediction Center (CPC) and the National Ocean and  
131 Atmospheric Administration (NOAA). This data has an original spatial resolution of 0.50 degrees  
132 (approximately 50 km), with interpolation to 12.5 km, and a temporal resolution of 6 hours. As for  
133 the air pollution data, we calculated the daily mean temporal resolution for each weather variable,  
134 and then we aggregated the data by the municipality.

135

### 136 **2.4. Statistical analysis**

137 A time-series analytic approach was applied to look at the relationship between short-term  
138 exposure to PM<sub>2.5</sub>, NO<sub>2</sub>, O<sub>3</sub>, and admissions with cardiovascular and respiratory outcomes. Our  
139 statistical approach was divided into three stages, as described below.

140 First, we used a distributed lag modeling framework to create two cross-basis functions,  
141 including one for the exposure (air pollution) and the other one for temperature – a potential  
142 confounder, as suggested by the literature (Di et al., 2018; Shi et al., 2021; Weinberger et al., 2021).  
143 For the effect of the air pollutant, we defined the cross-basis function with a linear term for the air  
144 pollution-hospitalization relationship, known as the distributed lag linear model (DLM). We  
145 accounted for the lagged effect of each air pollutant up to 15 days of lag (0-15 days) with a 4<sup>th</sup>-  
146 degree polynomial function.

147 For the temperature effect, we used a distributed lag non-linear model – DLMN (Gasparrini  
148 et al., 2010) to define a cross-bases function with a natural cubic spline with 5 degrees of freedom  
149 for the temperature exposure-response and then 2 lagged terms (lag strata), including lag 0 and lag



150 1-15. We considered the effects as constant within each stratum. The internal knots were placed  
151 by natural cubic spline equally spaced quantiles, while the boundary knots were located at the  
152 temperature range. Note that although we are using non-linear term to represent the distributed lag  
153 function for the confounder (temperature), the modeling framework of our study can be defined as  
154 a DLM, given that the effect of the exposure (air pollution) was included in the cross-basis function  
155 with a linear term.

156 In the second stage, we included the two cross-basis functions in a generalized conditional  
157 quasi-Poisson regression model. We fit only single-air pollutant models. The analyses were  
158 stratified by sex, age group (15-25, 26-35, 36-45, 46-55, 56-65, >65 years old), health outcome  
159 (respiratory, circulatory, and cardiorespiratory), and Brazilian regions (North, Northeast, Midwest,  
160 Southeast, and South).

161 Relative risks (RRs) and 95% confidence interval (CI) were derived to represent the  
162 association of every 10 units increases each air pollutant with the hospitalization for circulatory  
163 and respiratory diseases. We used  $(RR-1) \times 100\%$  to report the percentage increase in the health  
164 outcome.

165 To control for temporal trends within each municipality in the Poisson model, we used a  
166 time-stratified sampling to define strata based on the day of the week, month, calendar year, and  
167 the municipality of the time series. This allows the comparison of the exposure on the day of a  
168 health event on Monday in January 2008, for example, with exposures on all other Mondays in  
169 January 2008. Note that in this study design, each case period has 3 or 4 control periods. Also  
170 given the matching periods were close in time, the approach reduces the effects of confounding  
171 related to the seasonal trend by controlling for time-dependent risk factors, including the day of  
172 the week, season, and long-term trends by matching. Note that although we included the  
173 municipalities in the strata, we obtained the impact (RR) by region, not by municipality.

174 Finally, in the third stage, we applied meta-analysis to estimate a national RR. Specifically,  
175 we accounted for intra- and inter-region variability by applying regression meta-analysis with  
176 random effects. This approach assumes that the average effect size in the Brazilian population  
177 varies randomly from region to region (Viechbauer, 2010). Heterogeneity was examined using the  
178 I-squared ( $I^2$ ) statistic. A p-value  $>0.05$  and/or  $I^2 <50\%$  was considered homogeneous.

179 All computations of this study were run on the Google Cloud Platform. R software, version  
180 4.0.2, was used for all analyses. We used the R package “dlnm” (Distributed Lag Non-Linear

181 Models) for the distributed lag modeling framework and the R package “gmm” (Generalized  
182 Nonlinear Models) for the generalized conditional quasi-Poisson regression model.

183

## 184 **2.5. Sensitivity analysis**

185 We tested the robustness of our results by changing several modeling assumptions. First,  
186 we added in the primary Poisson model the adjustment for humidity, resulting in a model with the  
187 exposure (air pollution) represented by the cross-basis function (first stage), one confounder  
188 represented by the cross-basis function for the temperature (first stage), and humidity as the second  
189 confounder (without cross-basis function). Subsequently, we tested a model with one more  
190 confounder, precipitation (humidity + precipitation, both without cross-basis functions). We also  
191 analyzed the sensitivity of our primary model by removing the adjustment for the cross-basis  
192 function for temperature. Note that this was a simple unadjusted model, with only the exposure.  
193 Finally, we checked the robustness of our primary results by changing the cross-basis function for  
194 the air pollution-hospitalization relationship. Specifically, we defined a model with a non-linear  
195 term represented by a B-spline with 5 degrees of freedom for the lag structure.

196

## 197 **3. RESULTS**

### 198 **3.1. Hospital admission characteristics**

199 Table 1 provides the descriptive characteristics of these health events in Brazil during our  
200 study period. Our study population includes 23,791,093 hospital admissions for cardiorespiratory  
201 diseases in Brazil between 2008 and 2018. Among those, 53.1% are respiratory diseases, and  
202 46.9% are circulatory diseases. The number of male admissions was slightly higher for both  
203 respiratory and circulatory diseases, representing 52.7% and 50.2% of the total admissions,  
204 respectively. For the age groups, the largest proportion was for patients aged >65 years, with  
205 25.85% hospitalized due to respiratory diseases and 45.11% due to circulatory diseases. Figure 1  
206 shows Brazil's spatial distribution of respiratory and circulatory hospital admissions in the study  
207 period.

208 Table 1 – Descriptive characteristics of hospital admission events in Brazil, 2008-2018.

Health outcome	Age	Number of hospital admissions (%) <sup>1</sup>		
		Men	Women	All sex
<b>Respiratory hospital admissions</b>	15-25	382,347 (3.03)	394,697 (3.13)	777,044 (6.16)
	26-35	322,265 (2.55)	319,398 (2.53)	641,663 (5.08)
	36-45	324,760 (2.57)	309,308 (2.45)	634,068 (5.02)
	46-55	423,052 (3.35)	403,144 (3.19)	826,196 (6.55)
	56-65	565,438 (4.48)	507,324 (4.02)	1,072,762 (8.50)
	>65	1,630,513 (12.92)	1,632,364 (12.93)	3,262,877 (25.85)
	All ages <sup>2</sup>	6,646,890 (52.66)	5,975,080 (47.34)	12,621,970 (100)
<b>Circulatory hospital admissions</b>	15-25	125,708 (1.13)	140,488 (1.26)	266,196 (2.38)
	26-35	219,681 (1.97)	313,996 (2.81)	533,677 (4.78)
	36-45	445,269 (3.99)	572,156 (5.12)	1,017,425 (9.11)
	46-55	949,565 (8.5)	917,064 (8.21)	1,866,629 (16.71)
	56-65	1,412,606 (12.65)	1,135,610 (10.17)	2,548,216 (22.81)
	>65	2,516,053 (22.53)	2,522,032 (22.58)	5,038,085 (45.11)
	All ages <sup>2</sup>	5,611,256 (50.24)	5,557,867 (49.76)	11,169,123 (100)

209 Notes: <sup>1</sup> the percentages were based on the total number of hospital admissions in Brazil between 2008 and 2018,  
210 which for respiratory hospital admissions were 12,621,970 cases, and for circulatory admissions were 11,169,123  
211 cases. <sup>2</sup> this includes people under 15 years old.

212

213

### 214 3.2. Characteristics of the exposure and covariate variables

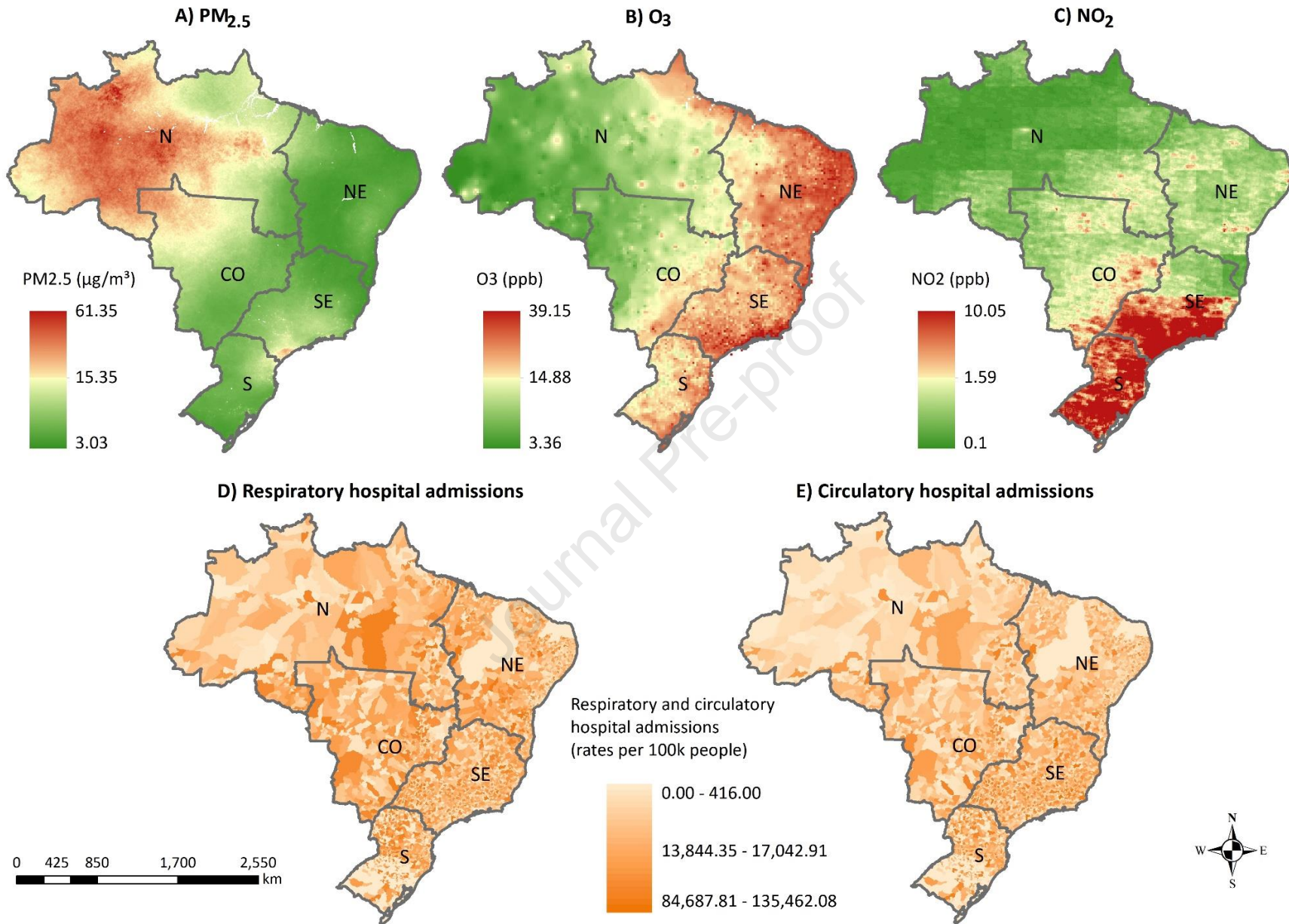
215 Table 2 shows the summary statistics for air pollution and weather variables. Figure 1  
 216 illustrates the nationwide concentrations of air pollutants in Brazil. The mean concentrations (with  
 217 the standard deviations) of PM<sub>2.5</sub>, NO<sub>2</sub>, and O<sub>3</sub> over the study period in Brazil were 10.01 (8.28)  
 218 µg/m<sup>3</sup>, 1.51 (1.48) ppb, and 20.07 (7.01) ppb, respectively. The mean PM<sub>2.5</sub> exceed the 2021 Air  
 219 Quality Guidelines (2021 AQG) from the World Health Organization - WHO (5µg/m<sup>3</sup> annual  
 220 mean). To compare the concentrations of NO<sub>2</sub> and O<sub>3</sub> with the 2021 AQG, we converted the  
 221 original unit of these pollutants (ppb) to the unit used in the 2021 AQG (µg/m<sup>3</sup>). The WHO  
 222 recommends the following conversion factors: for NO<sub>2</sub>, 1 ppb = 1.88 µg/m<sup>3</sup>; for O<sub>3</sub>, 1ppb = 1.96  
 223 µg/m<sup>3</sup>. After applying the conversion factors, the estimated mean of NO<sub>2</sub> and O<sub>3</sub> concentrations  
 224 were 2.83 µg/m<sup>3</sup> and 39.33 µg/m<sup>3</sup>, respectively, which did not exceed the 2021 AQG (10 µg/m<sup>3</sup>  
 225 annual mean for NO<sub>2</sub> and 60 µg/m<sup>3</sup> peak season for O<sub>3</sub>). For the weather variables, we estimated  
 226 a mean (with the standard deviation) of 23.59 (3.79)°C for temperature, 77.76 (13.64)% for relative  
 227 humidity, and 3.39 (8.16) mm/day for precipitation.

228  
 229 Table 2 – Summary statistics for air pollution and weather data in Brazil, 2008-2018.

Variable	Min	Q1	Mean	SD	Q3	Max
PM <sub>2.5</sub> (µg/m <sup>3</sup> )	3.03	5.05	10.01	8.28	11.85	61.35
NO <sub>2</sub> (ppb)	0.10	0.60	1.51	1.48	1.88	10.05
O <sub>3</sub> (ppb)	3.36	15.22	20.07	7.01	24.77	39.15
Temperature (°C)	1.25	21.60	23.59	3.79	26.23	34.92
Relative humidity (%)	20.00	69.75	77.76	13.64	88.50	100
Precipitation (mm/day)	0	0	3.39	8.16	3.00	486.00

230 Note: Minimum (Min), first quartile (Q1), standard deviation (SD), third quartile (Q3), and maximum (Max).

231



232

233 Figure 1 – Spatial variation of respiratory and circulatory hospital admissions (sum between 2008-2018) and air pollution concentrations  
 234 (average between 2008-2018) in Brazil. Note: the polygons in gray represent the Brazilian regions, including North (N), Northeast (NE), Midwest (CO), Southeast  
 235 (SE), and South (S).

### 236 3.3. Association between air pollution and hospital admissions

237 Figures 2-4 show the national average RR of hospital admissions associated with PM<sub>2.5</sub>,  
238 O<sub>3</sub>, and NO<sub>2</sub>, respectively. This is the overall cumulative effect of a 10-unit increase in each  
239 ambient air pollutant over 15 days of lag (summing all the contributions up to the maximum lag).  
240 We present in the Supplementary Materials a table with all the national average RR (from the  
241 primary analysis, for all sub-group analyses, pollutants, and health outcomes) along with the  
242 estimated heterogeneity test from the meta-analysis (Table S1).

243 The national meta-analysis for the whole dataset (without stratification by sex and age)  
244 showed that for every increase of PM<sub>2.5</sub> by 10 µg/m<sup>3</sup>, there is a 3.28% (95%CI: 2.61; 3.94) increase  
245 in the risk of hospital admission for respiratory diseases and a 1.86% (95%CI: 0.83; 2.89) increase  
246 in the risk of hospital admission for cardiorespiratory diseases. The associations were highly  
247 uncertain for circulatory hospital admissions (Figure 2). The subgroup analysis for sex and age  
248 indicated a positive association with respiratory diseases for almost all groups. Uncertain  
249 associations were found only for men 15-25 years old, men 46-55 years old, and women 15-25  
250 years old. For circulatory admissions, we observed positive associations only for men/women 36-  
251 45 years old and women 36-45 years old (Figure 2).

252 The national meta-analysis for O<sub>3</sub> (Figure 3) indicated a positive association with  
253 respiratory hospital admissions for eight sub-group analyses, mostly for people aged >45 years  
254 old, including men/women 46-55, men/women 56-65, men/women >65, men >65, women 36-45,  
255 women 46-55, women 56-65, and women >65. For circulatory diseases, the results were similar,  
256 with some difference that indicated positive associations for youngest population, including people  
257 aged 15-25 and 26-35 years old (Figure 3).

258 At the national level, the highest percentage increase in the risk of hospital admissions for  
259 respiratory diseases occurred when we used NO<sub>2</sub> as the exposure (Figure 4). The national meta-  
260 analysis for the whole dataset showed that a 10ppb increase in NO<sub>2</sub>, there is a 35.26% (95%CI:  
261 24.07; 46.44) and 16.90% (95%CI: 11.17; 22.64) increase in the risk of hospital admission for  
262 respiratory and cardiorespiratory diseases, respectively. For respiratory hospital admissions, this  
263 positive association persisted for several sub-group analyses, including the group with the oldest  
264 people, men >65 and women >65. There were no positive associations for the relationship between  
265 NO<sub>2</sub> and hospital admissions for circulatory diseases (Figure 4).



266 Similar to the figures 2-4 presented here, we show in the Supplementary Materials the  
267 charts of the RR (from the primary analysis) stratified by Brazilian regions (Figures S2-S10).  
268 Overall, when we looked at the PM<sub>2.5</sub> exposure, the North was the region with the greatest number  
269 of positive and consistent associations among the sub-group analyses for the three health  
270 outcomes, respiratory, circulatory, and cardiorespiratory hospital admissions (Figures S2-S4).  
271 Specifically, when we examined cardiorespiratory hospital admissions, our findings suggested a  
272 positive association with PM<sub>2.5</sub> in all sub-group analyses, except for three groups, including  
273 man/women, 46-55; men, 15-25; and men, 46-55 (Figure S4). For NO<sub>2</sub>, when we examined the  
274 whole dataset (without stratification by sex and age), we found positive associations with  
275 respiratory admissions and cardiorespiratory admissions in all Brazilian regions (Figures S5-S7).  
276 This positive association persisted across the regions when we stratified the analyses by men of all  
277 ages and women of all ages (Figures S5 and S7). In contrast, only the Midwest and Southeast  
278 regions presented a few positive associations when looking at the relationship between NO<sub>2</sub> and  
279 circulatory diseases (Figure S6). For O<sub>3</sub>, Midwest, Southeast, and South were the regions most  
280 impacted when we examined respiratory and cardiorespiratory hospital admissions (Figures S8  
281 and S10). Table S2 in Supplementary Materials shows all the RR values from the primary and  
282 sensitivity analysis for all sub-groups, pollutants, health outcomes, and regions.

283 Lag-response curves for a 10-unit increase in ambient air pollution in each region are  
284 shown in Figure 5. Lag-response curves of incremental cumulative effects are shown in Figure 6.  
285 For both lag-response curves (Figures 5 and 6), we show only the results for the respiratory hospital  
286 admissions, the most critical health outcome, according to the above results. Our results indicate  
287 that the lag-response curves are very heterogeneous among the regions and air pollutants (Figure  
288 5). For example, in the Southeast, for every increase of PM<sub>2.5</sub> by 10µg/m<sup>3</sup>, there is approximately  
289 0.3% increase in the risk of hospital admission due to respiratory diseases on the same day as the  
290 exposure, lag 0 (Figure 5). The exposure from about 7-10 days ago has an uncertain association  
291 (95% CI includes the value 1) with respiratory admissions. In contrast, in the Midwest region,  
292 PM<sub>2.5</sub> presented a positive association with respiratory admissions over 15 days of lag (Figure 5).

293 Regarding the cumulative relative risk (Figure 6), we can also observe heterogeneous lag-  
294 response curves among the regions and air pollutants. Overall, there is a consistent increased  
295 cumulative risk over 15 days of lag in all regions, except for O<sub>3</sub> in the North, Northeast, and South  
296 regions (Figure 6).

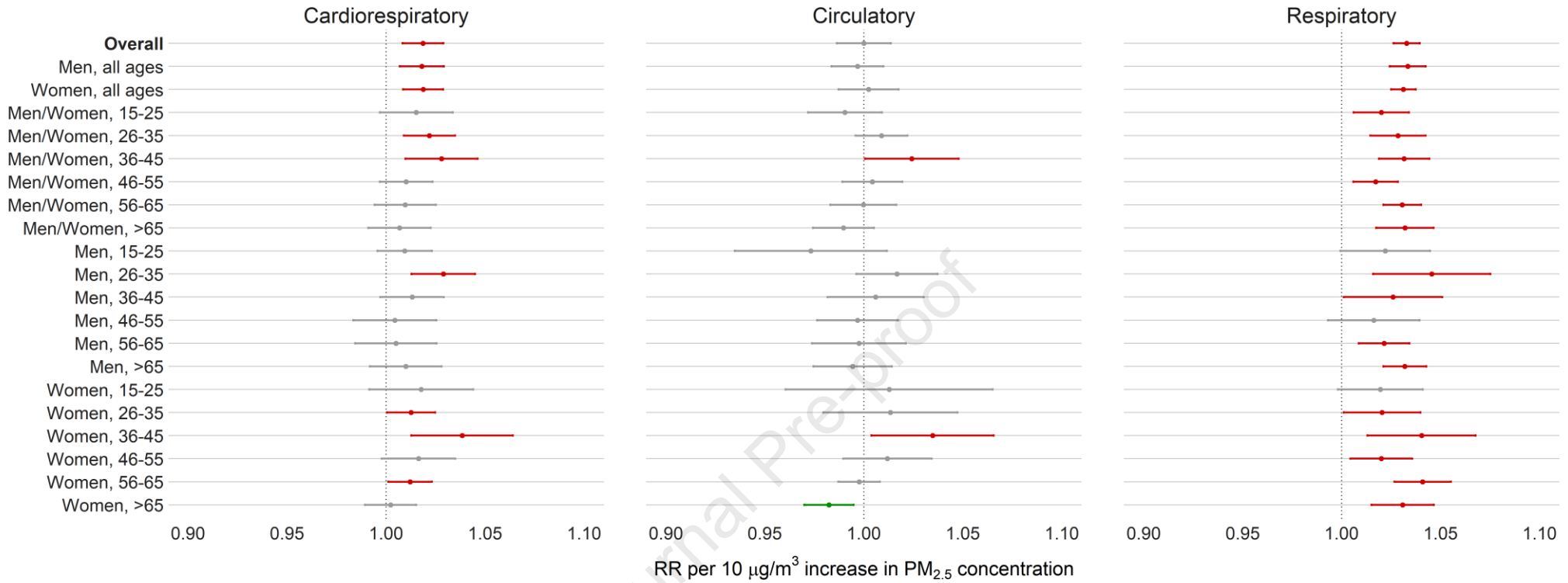


Figure 2 – Risk ratio (95%CI) per 10  $\mu\text{g}/\text{m}^3$  increase in  **$\text{PM}_{2.5}$  concentration** in Brazil (results from meta-analysis) stratified by health outcome, sex, and age group. This is the overall cumulative effect of a 10-unit increase in each ambient air pollutant over 15 days of lag (summing all the contributions up to the maximum lag).

Note: gray color represents the insignificant coefficients (which the RR includes the value 1), red color represents the significant positive associations, and green color represents the significant negative associations.



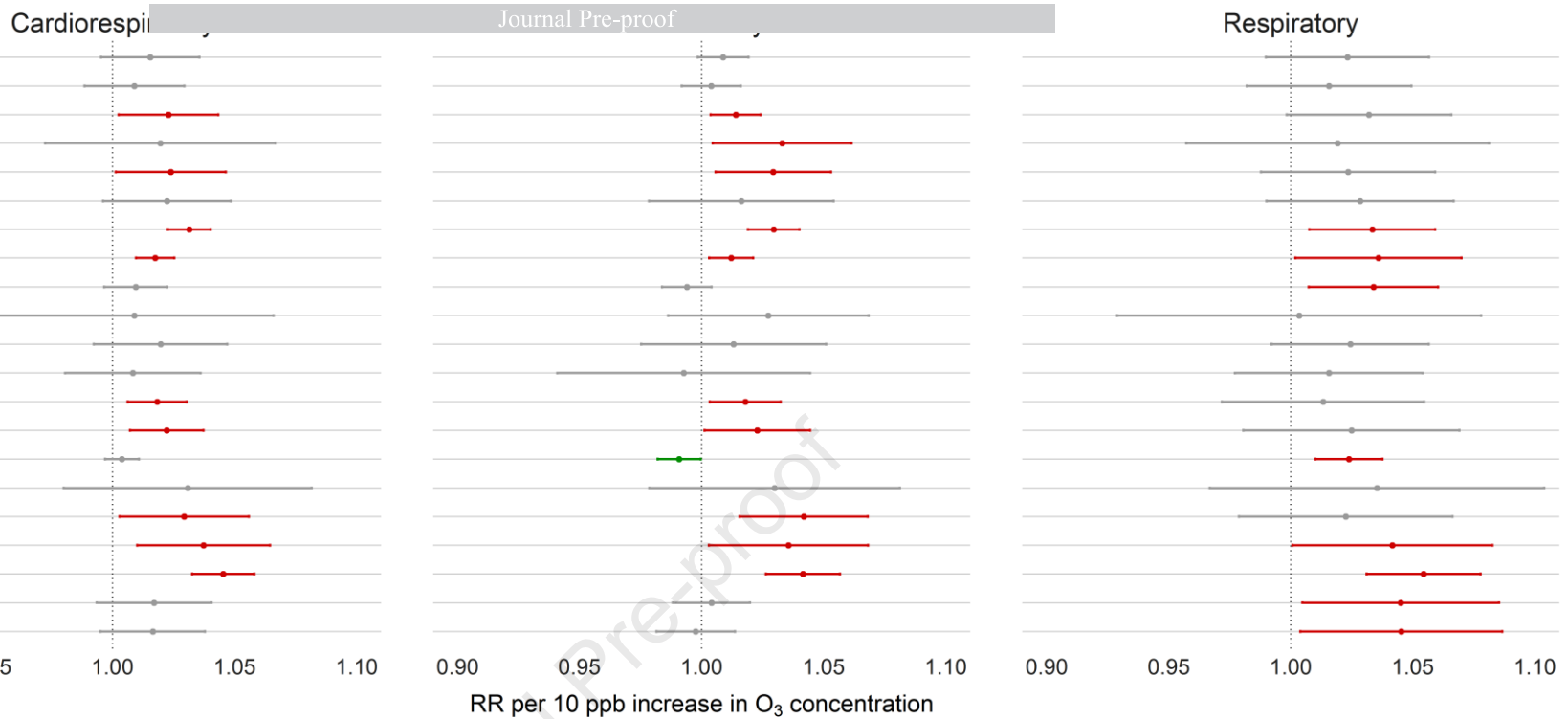


Figure 3 – Risk ratio (95%CI) per 10 ppb increase in **O<sub>3</sub> concentration** in Brazil (results from meta-analysis) stratified by health outcome, sex, and age group. This is the overall cumulative effect of a 10-unit increase in each ambient air pollutant over 15 days of lag (summing all the contributions up to the maximum lag).

Note: gray color represents the insignificant coefficients (which the RR includes the value 1), red color represents the significant positive associations, and green color represents the significant negative associations.

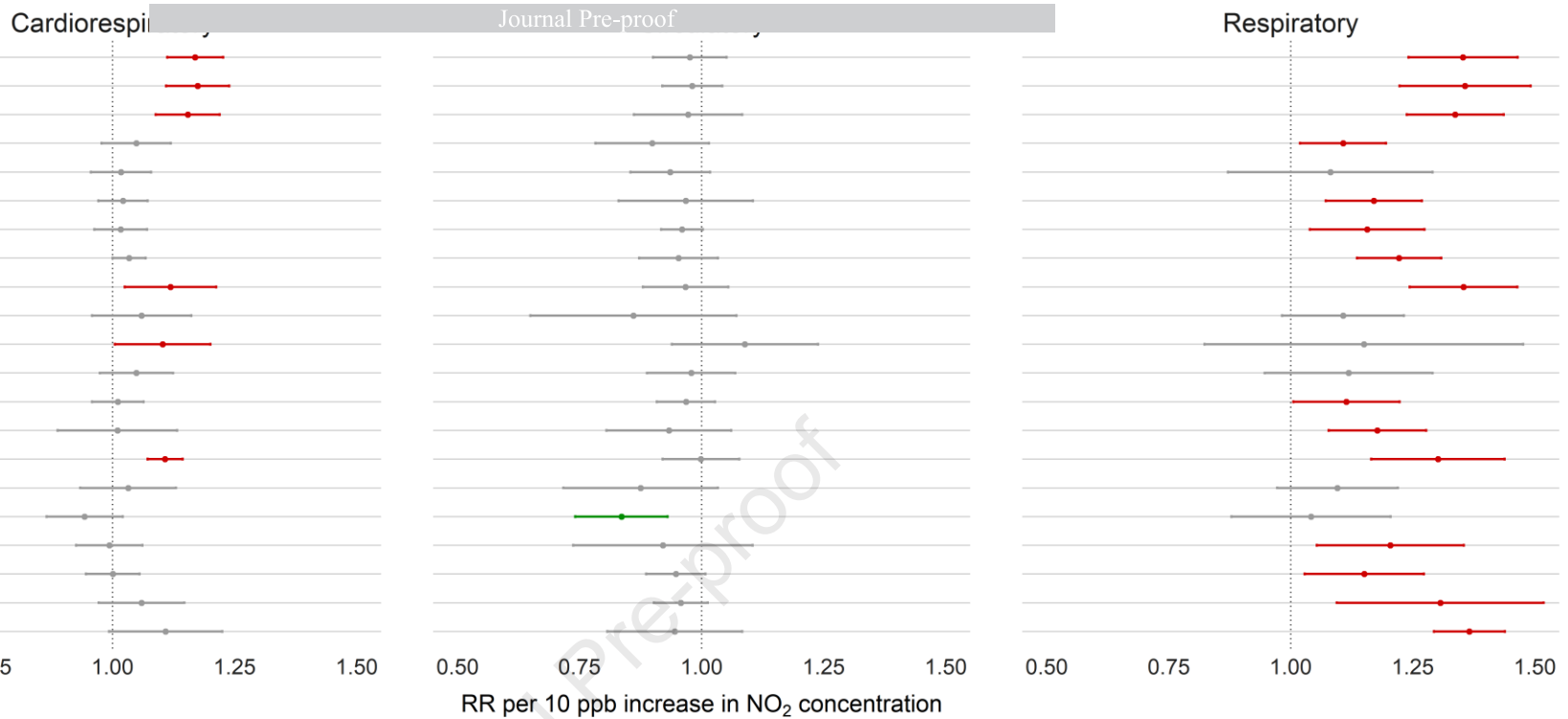


Figure 4 – Risk ratio (95%CI) per 10 ppb increase in **NO<sub>2</sub> concentration** in Brazil (results from meta-analysis) stratified by health outcome, sex, and age group. This is the overall cumulative effect of a 10-unit increase in each ambient air pollutant over 15 days of lag (summing all the contributions up to the maximum lag).

Note: gray color represents the insignificant coefficients (which the RR includes the value 1), red color represents the significant positive associations, and green color represents the significant negative associations.

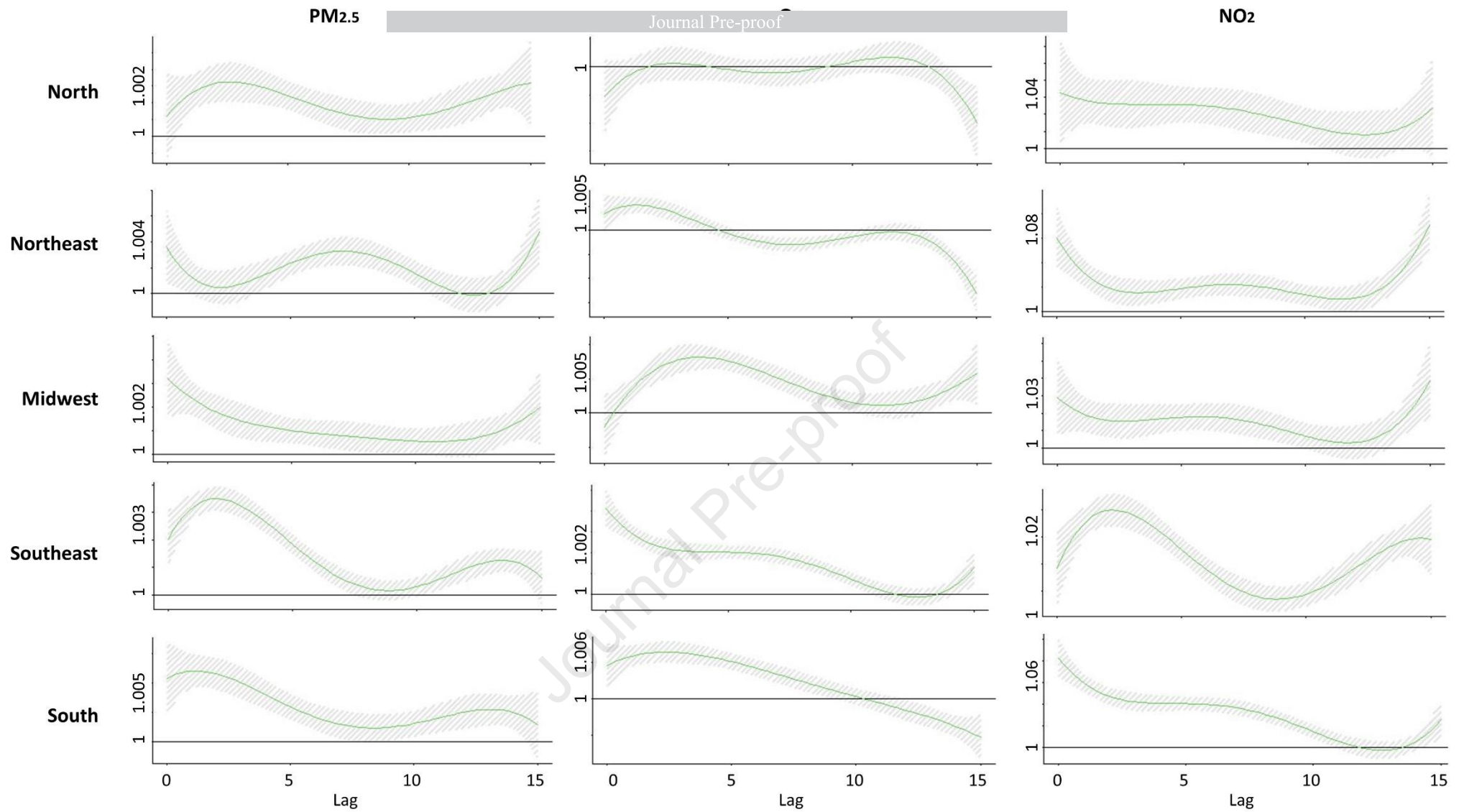


Figure 5 – Risk ratio (95%CI) per 10-unit increases in air pollution by lag for respiratory hospital admissions for the overall population. Charts are stratified by air pollutants and regions.

Note: the y-axis represents the risk ratio (RR).

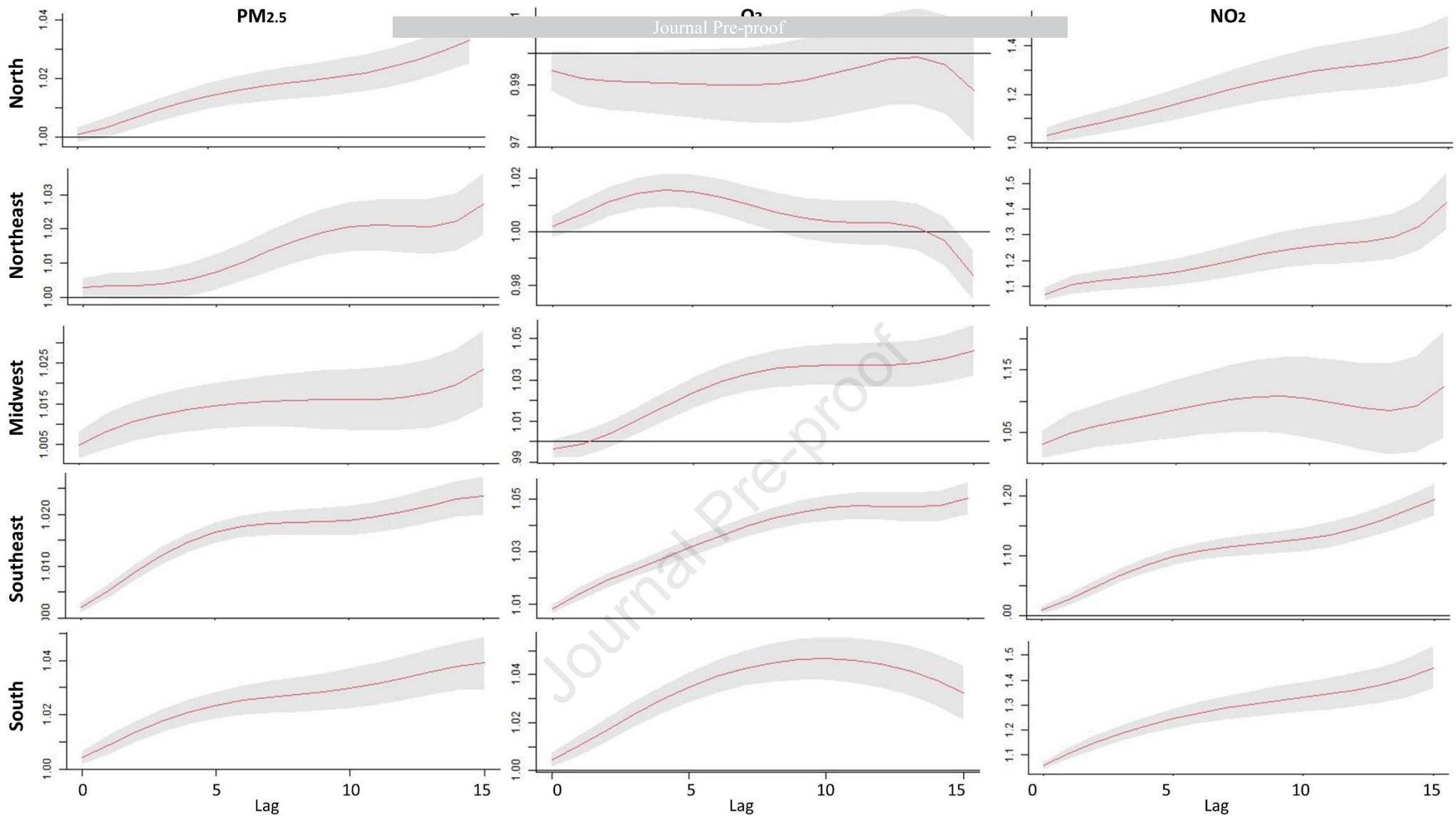


Figure 6 – Cumulative risk ratio (95%CI) per 10-unit increases in air pollution for respiratory hospital admissions for the overall population. Charts are stratified by air pollutants and regions.

Note: the y-axis represents the risk ratio (RR).

#### 338 4. DISCUSSION

339 To our knowledge, this is the largest and most comprehensive study to quantify the  
340 relationship between short-term exposures to ambient air pollution and hospital admissions in  
341 Brazil. We explored the effects of the three most impactful ambient air pollutants, PM<sub>2.5</sub>, O<sub>3</sub>, and  
342 NO<sub>2</sub> (previous studies in Brazil have only accounted for PM<sub>2.5</sub>). We included approximately 23  
343 million hospital admissions from all Brazilian municipalities, 5,572 municipalities (previous  
344 studies in Brazil have considered less than 2,000 municipalities). Our findings suggest a significant  
345 association of ambient air pollution (PM<sub>2.5</sub>, NO<sub>2</sub>, and O<sub>3</sub>) with respiratory and circulatory hospital  
346 admissions in Brazil. The statistical significance, the direction, and the effect size of this  
347 association varied substantially depending on the type of air pollutant, region, sex, and age group.  
348 This is consistent with results from previous Brazilian investigations (a limited number of studies,  
349 as presented in the introduction) at the national level and local Brazilian studies. We highlight that  
350 the comparison of our results with previous studies performed in Brazil that we present below  
351 should be done with caution due to the differences in the study design, model, and data among our  
352 study and the Brazilian literature. This comparison should be done based on the public health  
353 overview, particularly, looking at the evidence about the health effects risk associated with air  
354 pollution exposure.

355 A recent study at the national level in Brazil focused on the association between PM<sub>2.5</sub>  
356 (only this pollutant) and cancer hospitalizations (Yu et al., 2021), and found that with each 1µg/m<sup>3</sup>  
357 increase in two-year-average concentrations of PM<sub>2.5</sub>, the RR of hospitalization was 1.04 (95%CI:  
358 1.02; 1.07) for all-site cancers from 2002 to 2015 without sex and age differences. In our study,  
359 we estimated that for each 10µg/m<sup>3</sup> increase in PM<sub>2.5</sub> over 15 days of lag, the RR of respiratory  
360 and circulatory hospitalization were 1.032 (95%CI: 1.026; 1.039) without sex and age differences.  
361 We did not find significant associations for circulatory hospital admissions. Another Brazilian  
362 study at the national level looked at the association between hospital admissions and wildfire-  
363 related PM<sub>2.5</sub> and reported that a 10µg/m<sup>3</sup> increase in wildfire-related PM<sub>2.5</sub> was associated with a  
364 5.09% (95%CI: 4.73; 5.44) increase in respiratory hospital admissions and a 1.10% (95%CI: 0.78;  
365 1.42) increase in circulatory hospital admissions, over 0–1 day after the exposure (Ye et al. 2021).  
366 In our study, for every increase of PM<sub>2.5</sub> by 10 µg/m<sup>3</sup>, we estimated a 3.28% (95%CI: 2.61; 3.94)  
367 increase in the risk of hospital admission for respiratory diseases over 0-15 days after de exposure.  
368 In addition, Ye et al. (2021) found the highest attributable rate in the North, South, and Midwest

369 regions, with the consistent results for respiratory diseases across regions and inconsistent for  
370 circulatory diseases. Our results were similar, in which circulatory hospital admissions presented  
371 the highest number of mixed results (statistical significance, direction of the association, and the  
372 effect size) among the regions.

373 In terms of health outcomes, our results may be more comparable with local studies in  
374 Brazil. (Gouveia et al., 2017) used Poisson models to estimate the association between air pollution  
375 and hospitalizations for cardiorespiratory diseases in the Metropolitan region of São Paulo, Brazil.  
376 The authors found that the RR of respiratory admissions for an increase of  $10 \mu\text{g}/\text{m}^3$  in the levels  
377 of  $\text{PM}_{10}$  ranged from 1.011 (95%CI: 1.009; 1.013) to 1.032 (95%CI: 1.024;1.040) in their study  
378 area. Inconsistent results were found for circulatory diseases (Gouveia et al., 2017). In our study, we  
379 estimated an RR of 1.023 (95%CI: 1.019; 1.027) associated with respiratory admissions for every  
380 increase of  $\text{PM}_{2.5}$  by  $10 \mu\text{g}/\text{m}^3$  over 15 days of lag in the Southeast region, where São Paulo is  
381 located. In contrast, another local study was performed in a city in the state of Rio de Janeiro (also  
382 located in the Southeast region), in which the investigators estimated an increase of 3.84%  
383 (95%CI: 0.40; 7.40) in hospital admissions from circulatory diseases for people aged 65 or more  
384 associated with each increment of  $10 \mu\text{g}/\text{m}^3$  in  $\text{PM}_{10}$  (Oliveira et al., 2017). The authors did not find  
385 significant associations when they used  $\text{O}_3$  as their exposure. In our study, for the overall  
386 cumulative effect over 15 days of lag (summing all the contributions up to the maximum lag), we  
387 found positive associations with circulatory hospital admissions when we subtracted the analysis  
388 for people aged 65 or more in the Southeast region only for  $\text{NO}_2$  as exposure, with an estimated  
389 increase of 5.31% (95%CI: 1.94; 8.79) in hospital admissions from circulatory diseases. We did  
390 not find studies in the Brazilian literature looking at the association between  $\text{NO}_2$  and hospital  
391 admissions.

392 Regarding the regional heterogeneity of the impact of ambient air pollution exposure  
393 indicated by our results, we suggest that part of this variability can be explained by Brazil's  
394 cultural, social, behavioral, and environmental/geographical conditions. These conditions may  
395 determine the use of health services by affected people during episodes with poor air quality,  
396 including i) the perception and decision of each person to seek medical care after getting symptoms  
397 of cardiorespiratory diseases during wildfire events and ii) the spatial distribution of health care  
398 facilities - e.g., hospitals, clinics, outpatient care centers. In Brazil, the distribution of health  
399 facilities varies drastically by region. For example, there are many more health facilities per capita

400 in the South than in the Amazon region, including the North. In Brazil, equity in health services is  
401 still poorly distributed among the regions. This important condition needs further exploration in  
402 studies exploring health impacts from ambient air pollution. In addition, the modeled exposure  
403 accuracy may vary across regions of Brazil.

404 Our findings suggest positive associations (only from PM<sub>2.5</sub> and NO<sub>2</sub>) with circulatory  
405 hospital admissions when we subtracted the analysis for women before menopause (aged 26-45  
406 years old). While this effect is not expected, given that postmenopausal women may be at higher  
407 risk (Miller et al., 2007), increased hospitalization due to circulatory events in women aged 26-45  
408 years old may be due to pre-existing condition diseases, including women with congestive heart  
409 failure (Mann et al., 2002), chronic lung disease (Goldberg et al., 2001), frequent arrhythmias  
410 (Mann et al., 2002), and diabetes (Hart et al., 2015).

411 Our study has some strengths. First, our sample size includes more than 23 million hospital  
412 admissions nationwide over a period of 11 years. To our knowledge, this study has the largest  
413 sample size and the most extensive study period in Brazil. As we mentioned above, a limited  
414 number of previous studies in Brazil have only conducted regional analyses (only in the Southeast  
415 region). Second, we used a modeling approach that flexibly describe associations showing  
416 potentially non-linear and delayed effects in time series data. Environmental stressors, such as air  
417 pollution, frequently show effects that are delayed in time, requiring specific models that account  
418 for the time dimension of the exposure-response relationship. Third, we accounted for  
419 spatiotemporal trends in our model by using a time-stratified sampling to define strata based on  
420 the day of the week, month, calendar year, and the municipality of the time series. This approach  
421 reduces the effects of confounding related to the spatiotemporal trend (a common effect with some  
422 environmental stressors as well, such as air pollution exposure) by controlling for time-dependent  
423 risk factors within the municipalities, including the day of the week, season, and long-term trends  
424 by matching.

425 Our results, however, should be interpreted considering some limitations. First, given the  
426 presence of the individual perceptions and decisions to seek medical care after symptoms of  
427 cardiorespiratory diseases, there will be people who went to the hospital on the first day of the  
428 exposure, people who wait until symptoms become too severe and went to the hospital five days  
429 after the exposure, people who got acute symptoms but did not go to the hospitals etc. Therefore,  
430 given that we used a time-series framework, this issue may have affected our results. Second, there



431 is a possibility of some residual confounding bias, even after adjusting for spatiotemporal factors,  
432 meteorological conditions, and stratifying the analysis by sex and age. Third, the predicted  
433 concentration of ambient air pollution may have resulted in some exposure measurement error.  
434 Fourth, our study design does not capture the cause-effect between exposure to ambient air  
435 pollution and cardiorespiratory diseases. We only estimated associations between exposure to air  
436 pollution and hospital admissions due to cardiorespiratory diseases.

437

## 438 **5. CONCLUSIONS**

439 We have found significant associations between exposure to PM<sub>2.5</sub>, NO<sub>2</sub>, and O<sub>3</sub> and  
440 cardiorespiratory hospital admissions. Most of these associations were not consistent in the  
441 subgroup analysis by sex and age. Given the lack of air pollution studies on the national scale in  
442 Brazil, further investigations are essential to consolidate a body of local literature that may provide  
443 better support for policymakers with the objective of improving air quality and public health.

444

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449

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- This is the largest and most comprehensive study to quantify the relationship between air pollution and hospitalizations in Brazil.
- Our study population includes 23,791,093 hospital admissions for cardiorespiratory diseases.
- Our findings suggest a significant association of ambient air pollution with cardiorespiratory admissions in Brazil.
- The statistical significance, the direction, and the effect size of the associations varied substantially depending on the type of air pollutant, region, sex, and age group.
- This study is essential in Brazil, a region with a lack of control and monitoring policies leading to increased air pollution.

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**Declaration of interests**

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

The authors declare the following financial interests/personal relationships which may be considered as potential competing interests:

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