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4 **1 Two-way effect modifications of air pollution and air temperature on total**
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6 **2 natural and cardiovascular mortality in eight European urban areas**
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62
63 **Abstract**
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65 **Background:** Although epidemiological studies have reported associations between
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67 mortality and both ambient air pollution and air temperature, it remains uncertain whether the
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69 mortality effects of air pollution are modified by temperature and vice versa. Moreover, little
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71 is known on the interactions between ultrafine particles (diameter ≤ 100 nm, UFP) and
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73 temperature.
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75 **Objective:** We investigated whether the short-term associations of particle number
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77 concentration (PNC in the ultrafine range (≤ 100 nm) or total PNC ≤ 3000 nm, as a proxy for
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79 UFP), particulate matter ≤ 2.5 μm ($\text{PM}_{2.5}$) and ≤ 10 μm (PM_{10}), and ozone with daily total
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81 natural and cardiovascular mortality were modified by air temperature and whether air
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83 pollution levels affected the temperature-mortality associations in eight European urban areas
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85 during 1999-2013.
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88 **Methods:** We first analyzed air temperature-stratified associations between air pollution and
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90 total natural (nonaccidental) and cardiovascular mortality as well as air pollution-stratified
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92 temperature-mortality associations using city-specific over-dispersed Poisson additive models
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94 with a distributed lag nonlinear temperature term in each city. All models were adjusted for
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96 long-term and seasonal trend, day of the week, influenza epidemics, and population dynamics
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98 due to summer vacation and holidays. City-specific effect estimates were then pooled using
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100 random-effects meta-analysis.
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103 **Results:** Pooled associations between air pollutants and total and cardiovascular mortality
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105 were overall positive and generally stronger at high relatively compared to low air
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107 temperatures. For example, on days with high air temperatures ($>75^{\text{th}}$ percentile), an increase
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109 of 10,000 particles/ cm^3 in PNC corresponded to a 2.51% (95% CI: 0.39%, 4.67%) increase in
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111 cardiovascular mortality, which was significantly higher than that on days with low air
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121 57 temperatures (<25th percentile) [-0.18% (95% CI: -0.97%, 0.62%)]. On days with high air
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123 58 pollution (>50th percentile), both heat- and cold-related mortality risks increased.
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125 59 **Conclusion:** Our findings showed that high temperature could modify the effects of air
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128 60 pollution on daily mortality and high air pollution might enhance the air temperature effects.
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132 **Keywords:**

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134 63 Ultrafine particles; particulate matter; ozone; air temperature; mortality; effect modification
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139 65 **Highlights:**

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142 66 • High air temperature enhanced the mortality effects of UFP, PM_{2.5}, PM₁₀, and O₃
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144 67 • Heat-related mortality risks were higher at high levels of PM_{2.5}, PM₁₀, and O₃
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146 68 • Cold effects on mortality were stronger when PNC was high
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149 69 • First study to investigate the interaction between UFP and temperature on mortality
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180 **71 1. Introduction**
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183 72 Exposure to ambient air pollution has been identified as a leading contributor to the global
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185 73 disease burden which caused 4.5 million deaths in 2015 (Cohen et al. 2017). Meanwhile, a
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187 74 large number of epidemiological studies has shown adverse impacts of exposure to both high
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189 75 and low ambient air temperatures on mortality (Basu and Samet 2002; Curriero et al. 2002;
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191 76 Guo et al. 2014; Ma et al. 2014). Given the increasing concern regarding the health impacts
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193 77 of climate change, interest has grown recently in estimating the joint effects of air pollution
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195 78 and air temperature on health. However, little is known about the potential interaction
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197 79 between air temperature and air pollution, which is crucial for estimating their joint health
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199 80 effects.
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202 81 Meteorological conditions affect surface air quality by influencing emissions,
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204 82 atmospheric chemistry, and pollutant transport (Fiore et al. 2015). Especially, ground-level
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206 83 ozone (O₃) is formed by chemical reactions between nitrogen oxides and volatile organic
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208 84 compounds in the presence of sunlight and high temperature (Crutzen 1974; Sillman 1999).
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210 85 Thus, air pollution can be influenced by air temperature. In studies assessing air pollution
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212 86 health effects, air temperature is usually controlled for as a confounder rather than a modifier
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214 87 (Chen et al. 2013; Li et al. 2017). The potential effect modification of air pollution on
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216 88 mortality by air temperature has been largely neglected, until recently, in epidemiological
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218 89 studies (Stafoggia et al. 2008). On the other hand, air pollution may amplify people's
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220 90 vulnerability to the adverse effects of temperature (Gordon 2003) and could act as an effect
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222 91 modifier in the short-term effects of air temperature on mortality (Breitner et al. 2014; Ren et
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224 92 al. 2006). This effect modification of temperature health effects by air pollution may be of
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226 93 great importance to public health benefits because air temperature is expected to continue to
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228 94 rise over the 21st century under all emission scenarios (IPCC 2013), whereas air pollution
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230 95 can be reduced in a few decades to yield measurable improvements in public health (Breitner
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239 96 et al. 2009; Pope III et al. 2009). Thus, both directions of effect modification, hence the two-
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241 97 way effect modifications, matter for public health under a warming climate and changing air
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244 98 quality.

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246 99 Although a few studies have examined the modifying effect by air temperature on
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248 100 particulate matter (PM)- and O₃-associated mortality, results are inconsistent regarding: (1)
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250 101 the direction of the interaction: most studies reported stronger PM or O₃ effects on days with
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252 102 high air temperatures (Jhun et al. 2014; Kim et al. 2015; Li et al. 2011; Qian et al. 2008; Ren
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254 103 et al. 2008a; Stafoggia et al. 2008), whereas few also reported stronger air pollution effects on
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256 104 days with low air temperatures (Chen et al. 2013; Cheng and Kan 2012; Sun et al. 2015); (2)
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258 105 the significance of interaction: among 12 studies of PM effects on daily total nonaccidental
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260 106 mortality, only six found statistically significant interactions, five observed nonsignificant
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262 107 interactions, and one reported significance only in Southern Chinese cities (Li et al. 2017;
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264 108 Meng et al. 2012). In contrast, only a limited number of studies have evaluated the modifying
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266 109 effect of air pollution on air temperature-related mortality (Breitner et al. 2014; Li et al. 2015;
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268 110 Ren et al. 2006). PM was found as a significant effect modifier in the association between
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270 111 temperature and total and cardiovascular mortality in Brisbane, Australia (Ren et al. 2006)
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272 112 and Guangzhou, China (Li et al. 2015), but not in three cities of Bavaria, Germany (Breitner
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274 113 et al. 2014). However, these studies have important limitations in characterizing the complex
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276 114 interaction between air temperature and air pollution: first, their analyses were based on a
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278 115 single city analysis; second, they assumed a linear effect, a single lag, or a moving average
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280 116 lag structure for temperature, therefore simplifying to a great extent the nonlinear and delayed
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282 117 temperature-mortality dependencies (Gasparrini et al. 2015b).

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286 118 Epidemiological evidence on whether air temperature modifies the effect of ultrafine
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288 119 particles (UFP) and vice versa is lacking, mostly due to the unavailability of routinely
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290 120 collected relevant data. UFP are hypothesized to have a high and independent toxic potential
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298 121 due to their small size (<100nm), large active surface area, and their ability to penetrate into
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300 122 the pulmonary alveoli and to translocate in the circulation (Brook et al. 2010; HEI Review
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302 123 Panel on Ultrafine Particles 2013). Few epidemiological studies have reported a (weak)
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304 124 positive association between short-term UFP exposure and mortality (Atkinson et al. 2010;
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306 125 Breitner et al. 2011; Breitner et al. 2009; Lanzinger et al. 2016; Stafoggia et al. 2017).

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308 126 In the present study, we aimed to investigate the two-way effect modifications of air
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310 127 pollution (UFP, PM, and O₃) and air temperature on total (nonaccidental) and cardiovascular
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312 128 mortality in eight European urban areas. This study is the result of a collaborative effort
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314 129 among the Ultrafine Particles and Health (UF&HEALTH) Study Group in Europe (Stafoggia
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316 130 et al. 2017). The UF&HEALTH Study aimed to gather available data on UFP measures and
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318 131 mortality over a relatively long time period from cities across Europe to enlarge statistical
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320 132 power to detect weak associations (Samoli et al. 2016).

323 324 133 **2. Methods**

325 326 134 *2.1 Data collection*

327
328 135 Daily mortality, air pollution, and air temperature data during 1999-2013 were collected
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330 136 from eight European urban areas: Athens (Greece), Augsburg (Germany), Barcelona (Spain),
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332 137 Copenhagen (Denmark), Helsinki (Finland), Rome (Italy), Ruhr area (three adjacent cities
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334 138 including Essen, Mülheim, and Oberhausen, Germany), and Stockholm (Sweden)
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336 139 (Supplemental Information, Fig.S1). Detailed description of the study areas, including main
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338 140 sources of air pollution, are reported in the Supplemental Information, Text S1.

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340 141 Daily death counts of urban residents were provided by each participating center of the
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342 142 UF&HEALTH Study Group. Mortality data were classified into the following categories
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344 143 using the International Classification of Diseases, 9th revision (ICD-9) and the International
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346 144 Statistical Classification of Diseases and Related Health Problems, 10th revision (ICD-10):
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348 145 deaths from total natural (ICD-9 1-799 and ICD-10 A00-R99) and cardiovascular (ICD-9
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357 146 390-459 and ICD-10 I00-I99) causes. Respiratory mortality was not investigated because our
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359 147 previous study did not found associations of UFP and PM with respiratory mortality
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361 148 (Stafoggia et al. 2017). For total natural mortality, daily counts were also stratified by sex and
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363 149 age (0-74 years and 75 and above years). The two age groups (nonelderly vs. elderly) were
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365 150 used for analysis as previous studies suggested that the elderly are more vulnerable to the
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367 151 mortality risks of air pollution and air temperature (Anderson and Bell 2009; Bell et al. 2005;
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369 152 Hajat et al. 2007; Samoli et al. 2008).

372 153 Daily mean particle number concentration (PNC, as a surrogate for UFP (HEI Review
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374 154 Panel on Ultrafine Particles 2013)) was obtained from independent monitoring campaigns in
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376 155 each city. In all cities, one urban or suburban background PNC monitoring site was used,
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378 156 except for a traffic site in Rome. Due to different monitoring instruments used in different
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380 157 cities, PNC was measured in slightly different size ranges (Supplemental Information, Table
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382 158 S1). For Athens, Copenhagen, and Helsinki, PNC was available in the ultrafine range (≤ 100
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384 159 nm), in the other cities total PNC (≤ 3000 nm) was used as it is often assumed that particles in
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386 160 the ultrafine range dominated PNC (HEI Review Panel on Ultrafine Particles 2013). In each
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388 161 city, we further collected daily 24-h average PM with an aerodynamic diameter $\leq 2.5 \mu\text{m}$
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390 162 ($\text{PM}_{2.5}$) and $\leq 10 \mu\text{m}$ (PM_{10}) and daily maximum 8-h average O_3 concentrations from multiple
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392 163 stations of the local air quality monitoring networks. Daily concentrations were averaged
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394 164 from all valid monitoring stations in each city, which had at least 75% of the daily data for
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396 165 the study period. For details with regard to air pollution data collection we refer to the
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398 166 preceding publication (Stafoggia et al. 2017). As in previous studies, daily mean air
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400 167 temperature was used as the metric for temperature (Chen et al. 2016; Gasparrini et al.
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402 168 2015b). Data on daily mean air temperature were collected from local meteorological
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404 169 services or airport meteorological networks. Relative humidity was not collected since
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406 170 previous studies showed robust air temperature effects on daily mortality when additionally
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416 171 adjusting for relative humidity (Breitner et al. 2014; Gasparrini et al. 2015b; Guo et al. 2014).
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418 172 Influenza epidemics (a dummy variable denoting days with particularly high influenza
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420 173 episodes) were identified from national surveillance systems and hospitalization records.
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423 174 *2.2 Statistical Analysis.*

424 425 175 *2.2.1 Basic confounder model*

426
427 176 We used Poisson additive models with over-dispersion to estimate the city-specific
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429 177 associations between mortality and air pollutants or air temperature. Several confounders
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431 178 were included in the city-specific models: (1) natural cubic spline with eight degrees of
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433 179 freedom (*df*) per year to control for long-term and seasonal trend, (2) indicator variables for
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435 180 day of the week, (3) an indicator variable for influenza epidemics, (4) an indicator variable
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437 181 for population dynamics due to summer vacation and holidays (Stafoggia et al. 2017), and (5)
438
439 182 a penalized distributed lag nonlinear temperature term using marginal P-spline smoothers
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441 183 with dimension 7 for both exposure and lag spaces and a maximum lag of 21 days. The
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443 184 penalized distributed lag nonlinear temperature term was characterized as a cross-basis
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445 185 matrix, which had 42 (7×6) parameters for the bi-dimensional space of the exposure and lags.
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447 186 Penalization was implemented through a double varying penalty with a second-order
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449 187 difference penalty and a ridge penalty (Gasparrini et al. 2017). Because of the different lag
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451 188 periods for heat effect (within a few days) and cold effect (up to 3 or 4 weeks) (Anderson and
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453 189 Bell 2009; Gasparrini et al. 2015b), we applied a maximum of 21 lag days for temperature.
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456 457 190 *2.2.2 Air pollution effects stratified by air temperature*

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459 191 To examine effect modification by air temperature in each city, we categorized air
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461 192 temperature into three levels: high (>75th city-specific percentile), medium (25th-75th city-
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463 193 specific percentile), and low (<25th city-specific percentile). Consistent with prior studies
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465 194 (Chen et al. 2013; Jhun et al. 2014; Ren et al. 2008a), the 25th and 75th percentiles were used
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467 195 as temperature cut-offs. In addition, compared with other percentile cut-offs (5th and 95th, 10th
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475 196 and 90th, 15th and 95th, 20th and 80th), this percentile cut-offs could yield similar estimates but
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477 197 with narrower confidence intervals due to increased sample size in the low and high
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479 198 temperature levels (Chen et al. 2013; Jhun et al. 2014). After defining the basic confounder
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481 199 model, we introduced the interaction terms between air pollutant (PNC, PM_{2.5}, PM₁₀, and O₃
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483 200 in turn) and categorized air temperature at the same lag structure. Due to the multiple missing
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485 201 data in many of the air pollution series (Supplemental Information, Table S2), we could not
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487 202 compute averages over multiple days for air pollution. Based on our previous analysis
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489 203 (Stafoggia et al. 2017), we chose lag 6 for PNC and lag 1 for other pollutants. Heterogeneity
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491 204 among city-specific air pollution effects was assessed by the I² statistic from Cochran's Q test.
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493 205 Heterogeneity was considered to be significant if I² > 0.5, moderately significant if 0.25 < I²
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495 206 ≤ 0.5, and nonsignificant if I² ≤ 0.25 (Higgins et al. 2003).

498 207 *2.2.3 Air temperature effects stratified by air pollution concentrations*

500 208 For each city, we introduced an interaction term between the above mentioned penalized
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502 209 distributed lag nonlinear temperature term and an air pollutant strata indicator in the basic
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504 210 confounder model. To examine effect modification by air pollutants, we divided the air
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506 211 pollutants (PNC at lag 6, PM_{2.5}, PM₁₀, and O₃ at lag 1) into two levels: high (> city-specific
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508 212 median value) and low (≤ city-specific median value). Air pollution was categorized into two
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510 213 levels rather than three levels in order to ensure enough statistical power for the parameters in
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512 214 the cross-basis matrix of temperature and its interaction term with air pollution strata
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514 215 indicator. As the short-term effects of air pollutants are generally within several days (Bell et
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516 216 al. 2005; Samoli et al. 2008), we did not used the same cumulative lag structure (lag0-21) for
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518 217 air pollution and air pollution categories. To adjust for potential residual confounding, the air
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520 218 pollutant was also included as a linear continuous term in the model. The overall cumulative
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522 219 exposure-response curves for temperature and mortality were estimated along percentiles of
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524 220 the average temperature distribution in the eight European urban areas under study, with a
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534 221 minimum mortality temperature percentile between the first and the 99th percentiles as the
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536 222 reference temperature (Gasparrini et al. 2015b). Relative, city-specific temperature
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538 223 percentiles were used to characterize differences in temperature distributions and population
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540 224 acclimatization to temperature changes in cities with different climate conditions (Guo et al.
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542 225 2014; Jhun et al. 2014). Because the average temperature distributions were similar in
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544 226 different strata of PNC and PM but different in different strata of O₃ (Supplemental
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546 227 Information, Table S3), we constructed overall cumulative exposure-response relationships
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548 228 for each strata of air pollutants and represented these curves on a relative scale, along
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550 229 percentiles of the overall average temperature distribution. In addition, we calculated heat
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552 230 effects as cumulative mortality risk at the 99th percentile relative to the 90th percentile and
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554 231 cold effects as cumulative mortality risk at the 1st percentile relative to the 10th percentile.
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556 232 Since the 99th percentile (25.6 °C) is larger than the maximum value of temperature in low
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558 233 ozone levels, we calculated the heat effects in low ozone levels by comparing its maximum
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560 234 value (24.4 °C) with the 90th percentile (21.5 °C). The overall lag-response relationships for
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562 235 heat and cold effects across the lag period (0-21) were estimated separately.

566 236 City-specific effect estimates were pooled using univariate random-effects meta-analyses
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568 237 (Gasparrini et al. 2012). For temperature effects, city-specific coefficients for the cross-basis
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570 238 term were first pooled and then the pooled coefficients were used to reconstruct overall
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572 239 cumulative exposure-response associations on a relative scale using average temperature
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574 240 distribution percentiles (Gasparrini et al. 2015a). We tested the statistical significance of
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576 241 differences between the pooled estimates of the temperature or air pollutant strata by
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578 242 calculating the 95% confidence interval (CI) as $(\hat{Q}_1 - \hat{Q}_2) \pm 1.96\sqrt{(SE_{\hat{Q}_1})^2 + (SE_{\hat{Q}_2})^2}$, where
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580 243 \hat{Q}_1 and \hat{Q}_2 are the estimates, and $SE_{\hat{Q}_1}$ and $SE_{\hat{Q}_2}$ are their respective standard errors (Zeka et al.
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582 244 2006). We also tested the statistical significance of differences between the overall
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584 245 temperature-mortality associations at low and high air pollution levels using a multivariate
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593 246 Wald test based on the pooled reduced coefficients of the cross-basis matrix of temperature
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595 247 (Gasparrini et al. 2015a).

596 248 *2.3 Sensitivity analyses*

599 249 We performed several sensitivity analyses by changing the *df* (6-10 per year) for time
600 250 trend and using alternative maximum lag days for temperature (14 and 28 days). In addition,
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602 251 when analyzing modifications of the air pollution effects by air temperature, different cutoffs
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604 252 (20th/80th, 15th/85th, and 10th/90th) and lag days (lag 0 to lag 6) for temperature categories
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606 253 were also explored. Moreover, we fitted two-pollutant models by adding other co-pollutants
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608 254 one at a time to account for potential confounding from multiple exposures. Additionally, we
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610 255 explored whether differences in city-specific characteristics such as average temperature,
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612 256 temperature range, average air pollution level, and total number of population were
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614 257 associated with the estimated temperature-stratified air pollution effects. Using potential city-
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616 258 specific characteristics as additional meta-predictors, we then performed sensitivity analyses
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618 259 to pool the city-specific results using multivariate meta-regression models (Gasparrini et al.
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620 260 2012). Furthermore, we tested effect modification by sex and age group performing gender-
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622 261 and age-specific subgroup analyses. Besides, we compared the results of using UFP (3-100
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624 262 nm) with using total PNC (10-2000 nm) in Augsburg during 2004-2009. Finally, as Rome
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626 263 was previously found to dominate the pooled effects of PNC on mortality (Stafoggia et al.
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628 264 2017), we also checked the influence of Rome on the modification of air pollution effects by
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630 265 air temperature through removing it from the meta-analyses.

631 266 All analyses were performed with R software, version 3.2.1 (R Foundation for Statistical
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633 267 Computing, Vienna, Austria), using the packages mgcv (Wood 2011), dlnm (Gasparrini
634
635 268 2011), and mvmeta (Gasparrini et al. 2012).

636 269 **3. Results**

637 270 *3.1. Descriptive statistics*

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651
652 271 Table 1 summarizes daily mortality counts and cutoffs for air pollution and temperature
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654 272 strata in the eight European cities. Different research periods with available data on UFP
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656 273 measurements and mortality were investigated across different cities. During the study period,
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658 274 there were overall 742,526 total natural deaths in the eight cities, among which 39.3% were
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660 275 cardiovascular deaths. Daily total and cardiovascular mortality were highest in Athens and
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662 276 lowest in Augsburg. Median values of daily PNC ranged from 4,685 particles/cm³ in
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664 277 Copenhagen to 29,168 particles/cm³ in Rome. Cutoffs for both air pollutants and air
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666 278 temperature were generally higher in the Southern cities. The correlations of PNC with PM,
667
668 279 ozone, and air temperature, and correlations between PM and temperature were weak to
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670 280 moderate in each city (Supplemental Information, Fig.S2). On the contrary, ozone was
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672 281 moderately to strongly positively correlated with air temperature.
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675 282 *3.2. Air pollution effects modified by temperature*

676 283 Table 2 shows that the pooled effects of PNC, PM, and ozone on daily mortality varied by
677
678 284 temperature levels. Associations between increases in air pollutants and mortality were
679
680 285 generally stronger at high compared to low air temperatures. For example, a 10,000
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682 286 particles/cm³ increase in PNC at lag 6 was associated with percent increases in cardiovascular
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684 287 mortality of -0.18% (95% CI: -0.97%, 0.62%), 0.81% (95% CI: -1.92%, 0.32%), and 2.51%
685
686 288 (95% CI: 0.39%, 4.67%) at low, medium, and high air temperatures, respectively. The
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688 289 corresponding effect estimates on total mortality at each temperature level for a 10 µg/m³
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690 290 increase in PM_{2.5} were -0.46% (95% CI: -1.02%, 0.12%), 0.84% (95% CI: 0.05%, 1.63%),
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692 291 and 2.36% (95% CI: 0.11%, 4.65%). Nonsignificant or moderately significant heterogeneity
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694 292 ($I^2 \leq 0.5$) across different cities was observed for associations between mortality and PNC,
695
696 293 PM₁₀, and O₃, whereas significant heterogeneity ($I^2 > 0.5$) was found for associations
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698 294 between mortality and PM_{2.5} at high temperatures (Table 2 and Supplemental Information,
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700 295 Fig.S3-S6).
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711 296 *3.3. Air temperature effects modified by air pollutants*
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713 297 In the basic confounder model, the pooled air temperature-mortality associations were U-
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715
716 298 shaped and significant for both total natural and cardiovascular mortality (Fig.1). The lag-
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718 299 response relationships showed that heat effects were limited within the first week while cold
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720 300 effects lasted two to three weeks. No harvesting effect (deaths advanced by a few days) or
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722 301 mortality displacement was observed for both heat and cold effects.

723
724 302 Fig.2 shows the pooled estimates of the exposure-response relationship between air
725
726 303 temperature and total and cardiovascular mortality at low and high air pollution levels.
727
728 304 Associations between high temperatures and mortality were generally stronger at high PNC,
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730 305 PM, and O₃ levels. Estimates for low temperatures and mortality were much stronger at high
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732 306 PNC levels compared to low PNC levels, while were similar at PM and O₃ strata, with
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734 307 overlapping CIs. The results of the multivariate Wald test indicated evidence ($p < 0.05$) of
735
736 308 significant differences in the exposure-response curves for total natural mortality stratified by
737
738 309 PM and O₃ levels.

739
740 310 Table 3 reports the overall cumulative mortality risk of heat exposure (99th percentile
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742 311 relative to 90th percentile of air temperature) and cold exposure (1st percentile relative to 10th
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744 312 percentile of air temperature) by air pollutant strata. In general, both heat and cold effects on
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746 313 total and cardiovascular mortality were stronger at high air pollution levels. For example,
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748 314 heat exposure was associated with an increase in cardiovascular mortality by 19.02% (95%
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750 315 CI: -13.24%, 46.68%) at high PNC levels and 3.75% (95% CI: 0.29%, 7.33%) at low PNC
751
752 316 levels. Cold-related cardiovascular mortality risk was also higher at high PNC levels (16.23%;
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754 317 95% CI: 3.80%, 30.14%), compared to low PNC levels (2.00%; 95% CI: 0.16%, 3.88%).
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758 318 *3.4. Subgroup and sensitivity analyses*
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760 319 In population subgroup analyses, we did not find substantially different interactions
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762 320 between air temperature and PNC, PM, and O₃ on total natural mortality across age groups
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770 321 and sex (data not shown). Sensitivity analyses indicated that our results were robust when we
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772 322 changed *df* for time-trend (Supplemental Information, Fig.S7 and Fig.S8), used different
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774 323 percentile cutoffs of air temperature categories, and different lag periods for the air
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776 324 temperature effect (data not shown). Choosing different lag days for air temperature
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778 325 categories did not materially change the temperature-stratified air pollution effects on
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780 326 mortality (Supplemental Information, Fig.S9). After adjustment for co-pollutants, the pattern
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782 327 of effect modification on air pollution-related mortality by air temperature did not change
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784 328 substantially (Supplemental Information, Fig.S10). The effects of PNC on mortality across air
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786 329 temperature levels decreased after adjustment for PM_{2.5} but remained similar when
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788 330 controlling for PM₁₀ and ozone. Estimates of PM-related mortality across air temperature
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790 331 levels were robust when we controlled for PNC and ozone. Effect modification of ozone-
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792 332 related mortality by air temperature persisted after adjustment for PNC and PM. When we
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794 333 considered potential predictors (average temperature, temperature range, and population) of
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796 334 the city-specific risk estimates (Supplemental Information, Fig.S11), we found similar
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798 335 temperature-stratified air pollution effects (Supplemental Information, Fig.S12) and air
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800 336 pollution-stratified temperature effects (Supplemental Information, Fig.S13). Using UFP
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802 337 instead of total PNC generated similar results in Augsburg (Supplemental Information,
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804 338 Fig.S14). When we excluded Rome from the meta-analyses, the pooled effect modification of
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806 339 PNC- and PM-related cardiovascular mortality risks by high temperatures became
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808 340 nonsignificant, whereas effect modification of PM_{2.5}-related total natural mortality by high
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810 341 temperatures remained statistically significant (data not shown).

815 342 **4. Discussion**

818 343 To the best of our knowledge, this is the first time-series study to examine the interactions
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820 344 between UFP and air temperature on total natural and cardiovascular mortality. Our multi-
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822 345 city analyses in eight European urban areas showed that high temperatures could significantly

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829 346 enhance the effect of PNC on cardiovascular mortality, the effects of PM_{2.5} and PM₁₀ on total
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831 347 natural and cardiovascular mortality, and the effects of O₃ on total natural mortality.
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833 348 Furthermore, our results showed that the air temperature effects on mortality were greater at
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835 349 high air pollution levels. Significant effect modification was found on heat-related total
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837 350 natural mortality by PM_{2.5}, PM₁₀, and O₃, and on cold-related total natural and cardiovascular
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839 351 mortality by PNC.
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841 352 *4.1 Effect modification of air pollution effects by temperature*

842 353 We found stronger PM effects on mortality on days with high air temperatures. Similarly,
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844 354 high temperatures were found to enhance the acute effect of PM on mortality in Australia
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846 355 (Ren and Tong 2006), China (Li et al. 2011; Meng et al. 2012; Qian et al. 2008; Qin et al.
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848 356 2017), South Korea (Kim et al. 2015), and Europe (Katsouyanni et al. 2001; Pascal et al.
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850 357 2014; Shaposhnikov et al. 2014; Stafoggia et al. 2008). In the present analysis, an increase of
851
852 358 10 µg/m³ in PM₁₀ was associated with 0.03% (95% CI: -0.32%, 0.38%), 0.28% (95% CI:
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854 359 0.01%, 0.55%), and 0.93% (95% CI: 0.31%, 1.55%) increase of total natural mortality at low,
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856 360 medium, and high temperatures. Our results were consistent with a recent meta-analysis,
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858 361 which reported a 0.19% (95% CI: -0.01%, 0.40%), 0.31% (95% CI: 0.21%, 0.42%) and 0.78%
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860 362 (95% CI: 0.44%, 1.11%) increase in total natural mortality per 10 µg/m³ increase in PM₁₀ at
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862 363 study-specific low, medium, and high temperatures (Li et al. 2017). Moreover, in our study
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864 364 we observed a high heterogeneity of the PM_{2.5} effects between the cities and therefore our
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866 365 results should be regarded with caution.
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871 366 In accordance with our PM analysis, we also found stronger UFP effects on daily
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873 367 mortality on days with high temperatures. However, the effect modification was only
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875 368 significant for cardiovascular mortality. Evidence from very few studies on the seasonal
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877 369 association between PNC and mortality indicate that UFP effects may be larger in the warm
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879 370 season (Meng et al. 2013; Stafoggia et al. 2017), which provides support for our findings.
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888 371 Effect estimates were robust after adjustment for PM₁₀ and O₃, but weaker after adjustment
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890 372 for PM_{2.5}. In contrast, the temperature-stratified PM_{2.5} effects on mortality remained robust
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892 373 after adjustment of co-pollutants, which suggests independent effects of PM_{2.5}. This
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894 374 contrasting effects indicates some residual confounding in PNC effects due to co-exposure to
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896 375 PM_{2.5}. However, the contrasting results between PNC and PM_{2.5} should be interpreted with
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898 376 caution due to different size fractions of PNC measured in different cities. Except for Athens,
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900 377 Copenhagen, and Helsinki, total PNC (≤ 3000 nm) rather than PNC at ultrafine range (≤ 100
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902 378 nm) was measured (Supplemental Information, Table S1). In previous studies measuring UFP
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904 379 at ultrafine range (≤ 100 nm), the mortality effects of UFP remained similar when adjusting
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906 380 for PM_{2.5} (Lanzinger et al. 2016) or mass concentration metrics (Breitner et al. 2011).

909 381 City-specific effect modification of PNC effects by temperature showed different patterns,
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911 382 where largest effects occurring at high temperatures were observed in Athens, Augsburg,
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913 383 Barcelona, Helsinki, and Rome, but not in Copenhagen, Ruhr area, and Stockholm
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915 384 (Supplemental Information, Fig.S3). This difference may be due to different source
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917 385 contributions to UFP in different cities. A previous study evidenced that in Northern and
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919 386 Central EU cities PNC and black carbon (BC) had a similar hourly pattern, whereas in
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921 387 Southern EU cities, maximum PNC occurred at midday with minimum BC levels due to
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923 388 midday nucleation episodes (Reche et al. 2011). To quantify the sources and processes
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925 389 contributing to UFP, it can be segregated into two components based on the high correlation
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927 390 between BC and PNC: N1, the primary emission of vehicle exhaust, and N2, the newly
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929 391 formed secondary origin from mostly nucleation processes and other low BC-bearing UFPs
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931 392 from different sources (Brines et al. 2015; Cheung et al. 2011; Rodríguez and Cuevas 2007).
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933 393 Short-term effects of UFP on daily mortality are affected by different origins of UFP. A
934
935 394 recent study in three Spanish cities found an association of daily mortality with N1 but not
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937 395 with N2 in Barcelona and Santa Cruz de Tenerife, which were influenced by traffic emissions,
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947 396 whereas an association with N₂ was observed in an industrial city Huelva (Tobías et al. 2018).
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949 397 Thus, different source contributions of UFP in our eight EU cities may lead to different
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951 398 effects of PNC on daily mortality. Further studies with both PNC and BC measurements are
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953 399 need to differentiate modification effects of primary and secondary UFP on health by air
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955 400 temperature. Furthermore, city-specific modified PNC effects by temperature on total
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957 401 mortality were not fully explained by those effects on cardiovascular mortality. This suggests
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959 402 that PNC may have effects on other causes of deaths.
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962 403 A small number of studies have examined the modifying effect of air temperature on
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964 404 ozone-related mortality and the results are inconsistent (Li et al. 2017). In line with our
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966 405 findings, significant effect modifications of the association between O₃ and mortality with
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968 406 stronger effects on warmer days were found in the U.S. (Jhun et al. 2014; Ren et al. 2008a)
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970 407 and France (Pascal et al. 2012). On the contrary, stronger O₃ effects on colder days were
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972 408 observed in several cities in China (Chen et al. 2013; Cheng and Kan 2012; Liu et al. 2013).
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974 409 This difference may be likely due to inadequate control of cold effects in these studies by
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976 410 using short lags for temperature in the ozone-mortality association. A previous study in 21
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978 411 East Asia cities demonstrated that adjusting only for short lags of temperature could result in
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980 412 higher ozone effect estimates in winter than in summer (Chen et al. 2014).
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983 413 *4.2 Effect modification of temperature effects by air pollution*

984
985 414 Effect modification by air pollution on air temperature-mortality relationships has been
986
987 415 barely investigated. We observed higher heat- and cold-related mortality risks at high air
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989 416 pollution levels, with significant effect modification by PM_{2.5}, PM₁₀, and O₃ on heat-related
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991 417 mortality risks and by PNC on cold-related mortality risks (Table 3). Similar findings on
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993 418 PM₁₀ and O₃ were obtained by time-series studies conducted in Guangdong, China (Li et al.
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995 419 2015), Brisbane, Australia (Ren et al. 2006), 95 U.S. communities (Ren et al. 2008b), Berlin,
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997 420 Germany, and Lisbon, Portugal (Burkart et al. 2013), and three cities of Bavaria, Germany
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1006 421 (Breitner et al. 2014). Another study using a case-crossover design also reported larger heat
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1008 422 effects on mortality at high PM₁₀ concentrations in Rotterdam, The Netherlands (Willers et al.
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1010 423 2016). No prior investigations have assessed the modifying effect of short-term exposure to
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1013 424 PNC and PM_{2.5} on temperature-mortality associations.

1014 1015 425 *4.3 Plausible biological mechanism*

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1017 426 Although the underlying biological mechanism of effect modification of air pollution and
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1019 427 temperature on mortality is not fully understood, several hypotheses have been proposed.
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1021 428 Firstly, PM, O₃, and air temperature may have synergistic effects on cardiovascular system as
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1023 429 they have common pathophysiological pathways. Air temperature changes (higher or lower)
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1025 430 are associated with increased blood viscosity and coagulability, elevated cholesterol levels,
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1028 431 and inflammatory responses (Keatinge et al. 1986; Schneider et al. 2008). Increased UFP and
1029
1030 432 PM can also cause increased blood pressure and platelet aggregation, systemic oxidative
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1032 433 stress and inflammation (Brook et al. 2010; R ckerl et al. 2011). In addition, both airborne
1033
1034 434 particles and temperature were associated with changes in heart rate and repolarization
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1036 435 parameters among myocardial infarction survivors (Hampel et al. 2010). On the other hand,
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1038 436 ozone at high temperatures may impair fibrinolysis, thus reducing the efficiency of
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1040 437 preventing clot formation and clearance (Kahle et al. 2015). Second, high temperatures could
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1042 438 increase thermoregulatory stress and alter the physiological response to toxicants, leading to a
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1044 439 higher susceptibility to air pollution effects (Gordon 2003). Third, population exposures to air
1045
1046 440 pollution might increase during the warm season (Meng et al. 2013) as people tend to go
1047
1048 441 more outside and to keep windows open and at the same time the chemical composition of
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1050 442 UFP (Kim et al. 2002) and PM (Bell et al. 2007) could vary by season. In addition, secondary
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1052 443 UFPs formed from mostly nucleation events contributed as a major component of UFP in
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1054 444 Australian and European cities (Brines et al. 2015; Salma et al. 2014). Because nucleation
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1056 445 events generally occurred at midday with high temperature and low levels of nitrogen oxides
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446 (Brines et al. 2015), source contribution of UFP may greatly differ at low and high
447 temperatures. Seasonal variations in both chemical composition and source contribution of
448 UFP may affect its toxicity, which was observed to be higher in the summer (Baldauf et al.
449 2016).

450 *4.4 Strengths and limitations*

451 The eight European cities with PNC measurements offer advantages for the study of the
452 interactions between UFP and air temperature on daily mortality for the first time to our
453 knowledge. Furthermore, this study benefits from analyses on different particle sizes (UFP,
454 PM_{2.5}, and PM₁₀) and the potential synergistic role of temperatures. Another main strength of
455 this study is the multi-city design with standardized protocols for health data collection
456 covering a wide range of locations in Europe with different climates, which can provide
457 robust results and may avoid potential publication bias that commonly occur in single-city
458 studies. Moreover, disentangling interactions between the air pollution and air temperature on
459 health is challenging in part because of their different lag structures and a different shape of
460 their exposure-response functions (Zanobetti and Peters 2015). In the present analysis on
461 effect modification by air pollutant, rather than using a linear, single lagged or moving
462 averaged temperature term, we applied a distributed lag nonlinear temperature term, which
463 captures the complex non-linear and lagged dependencies in both the exposure-response and
464 lag-response associations (Gasparrini et al. 2015b). In the interaction term, this distributed lag
465 nonlinear temperature term was added together with a linear single lagged air pollution strata.
466 Thus, our models characterizing interactions with different lag structures and different
467 exposure-response functions may better assess the complex interplay between air pollutants
468 and air temperature on daily mortality.

469 Several limitations should be acknowledged in this study. First, there were potential exposure
470 measurement errors because we used measured air pollution and air temperature at fixed

1122
1123
1124 471 outdoor monitoring stations. This measurement error may be especially relevant to UFP as it
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1126 472 is known to have a high spatial variation within cities (HEI Review Panel on Ultrafine
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1128 473 Particles 2013). However, this concern was lessened to some extent as we analyzed the
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1130 474 temporal variations in time-series models and the temporal correlations across different sites
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1132 475 within a city were generally high (Cyrus et al. 2008). Second, different air pollution
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1134 476 measurement instruments were used and slightly different size fractions of PNC were
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1136 477 collected in different cities (Stafoggia et al. 2017), which might limit the direct comparison
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1138 478 among cities and introduce differential exposure measurement errors. Third, the UFP
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1140 479 measurements in Rome were influenced by traffic and had much higher particle number
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1142 480 concentrations, which may increase the statistical power and lead to the dominating role of
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1144 481 Rome in the pooled PNC effects (Stafoggia et al. 2017). Moreover, the multiple missing data
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1146 482 in air pollution measurements prevented us from conducting a sensitivity analysis using the
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1148 483 same cumulative lag structure for air temperature and air pollutants in assessing their
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1150 484 interactions. Furthermore, due to power issue we did not examine whether the observed effect
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1152 485 modifications varied by season. Further study is warranted to investigate the seasonal
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1154 486 interactions between air pollution and air temperature. Another limitation is that by testing
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1156 487 multiple air pollutants, temperature, and total and cardiovascular mortality, the possibility
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1158 488 that some of the observed significant effect modifications might occur by chance cannot be
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1160 489 fully excluded. In addition, our results might not be generalized to health impact assessments
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1162 490 in another region with different basic health status and air pollution compositions
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1164 491 (Krzyzanowski et al. 2002).

1165 492 **5. Conclusion**

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1167 493 Overall, our findings showed that the association between daily total natural and
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1169 494 cardiovascular mortality and air pollution (UFP, PM_{2.5}, PM₁₀, and ozone) was modified by air
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1171 495 temperature and vice versa. Results therefore suggest that interactions between air pollution

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496 and air temperature should be considered to assess their joint health effects. Our findings
497 point to the importance of understanding and reducing the health burdens attributable to
498 ambient air pollution and air temperature in the context of climate change. Further studies are
499 needed to investigate the effect modification of air pollution and air temperature using
500 morbidity data (i.e. hospitalization, emergency room visits) to get a more comprehensive
501 knowledge of the air temperature-pollution interaction.

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508 **Conflict of interest**

509 The authors declare no conflicts of interests.

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744 **Figure Legends**

745 **Fig.1.** Overall cumulative exposure-response relationships and lag-response relationships
746 between air temperature and mortality with 95% CIs. The vertical lines in (A) and (B)
747 represent the 1st, 10th, 90th, and 99th percentiles of the air temperature distribution. The y-
748 axis in (A) and (B) represents the relative risk of air temperature on daily mortality compared
749 with the minimum mortality temperature; in (C) and (D) represents the relative risk of heat
750 effect (99th percentile vs. 90th percentile) on daily mortality; and in (E) and (F) represents
751 the relative risk of cold effect (1st percentile vs. 10th percentile) on daily mortality.

752 **Fig.2.** Modified overall cumulative air temperature-mortality associations by air pollution
753 with 95% CIs. Blue lines represent for low air pollution level (concentration below median
754 value) and red lines represent a high air pollution level (concentration above median value).
755 The vertical lines represent the 1st, 10th, 90th, and 99th percentiles of the air temperature
756 distribution. The y-axis represents the relative risk of temperature on daily mortality
757 compared to the minimum mortality temperature. P value is the result of significance test
758 between air pollution levels, based on a multivariate Wald test of the pooled reduced
759 coefficients of the temperature effects at low and high air pollution levels.

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761 **Tables**

762 **Table 1.** Descriptive statistics for study period, daily deaths, and cutoffs for air pollutants and
763 air temperature in eight European cities.

	Helsinki	Stockholm	Copenhagen	Ruhr Area	Augsburg	Rome	Barcelona	Athens
<i>Study period</i>	2001-2010	2001-2005, 2008-2010	2001-2010	2009-2013	1999-2009	2001-2010	2005-2010	2008-2010
<i>Daily mortality (mean ± sd)</i>								
Total natural	18±5	39±7	26±9	32±6	8±3	58±10	41±8	80±12
Cardiovascular	7±3	16±4	8±4	12±4	4±2	24±6	13±4	36±7
<i>Air pollutants percentiles (median)</i>								
PNC (particles/cm ³)								
Min	793	2113	503	1513	2243	2295	1926	582
25th	4658	6210	3285	7523	7184	19384	13807	3775
50th	6636	8150	4685	9740	9977	29168	18696	5599
75th	9868	10910	6382	12331	13728	44202	24663	8696
Max	38761	44170	21260	28905	44755	139640	50929	36024
PM _{2.5} (µg/m ³)								
Min	0	1	0	4	1	0	2	5
25th	5	5	9	10	10	13	15	16
50th	7	7	11	14	15	18	20	20
75th	11	9	15	22	21	24	28	26
Max	57	37	70	128	126	73	104	63
PM ₁₀ (µg/m ³)								
Min	1	3	2	4	4	7	4	7
25th	10	9	14	14	19	26	23	24
50th	13	12	18	19	29	33	32	32
75th	19	17	23	28	40	44	43	42
Max	116	82	100	131	169	174	190	291
O ₃ (µg/m ³)								
Min	2	5	0	3	3	1	1	10
25th	48	51	26	36	45	42	41	49
50th	62	64	33	55	67	75	65	71
75th	76	78	40	75	93	100	83	91
Max	159	129	77	196	190	199	142	138
<i>Air temperature percentiles (°C)</i>								
Min	-24.2	-17.9	-8.1	-11.0	-12.9	-0.4	0	-1.8
1st	-18.0	-11.3	-4.5	-4.6	-7.8	2.6	1.7	4.9
10th	-6.5	-2.9	0.4	1.6	-0.9	6.6	6.4	10.0
25th	-0.7	1.2	4.3	6.4	3.6	10.2	9.5	13.8
50th	5.7	7.0	9.4	11.5	10	15.6	14.7	18.3
75th	13.7	13.9	15.2	16.7	16.2	21.9	19.6	25.5
90th	17.9	17.7	18.7	19.9	20.2	25.8	22.7	29.3
99th	22.8	22.3	22.9	25.2	24.9	28.9	25.7	32.2
Max	26.6	25.1	25.5	29.6	27.7	31	28.3	33.6

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765 **Table 2.** Percent increase (95% CI) in daily total natural and cardiovascular mortality
 766 associated with a 10,000 particles/cm³ increase in PNC or a 10 µg/m³ increase in PM_{2.5}, PM₁₀,
 767 and O₃ at different temperature levels.

Pollutant	Temperature levels ^a	Total natural		Cardiovascular	
		% increase	I ²	% increase	I ²
PNC_lag6	Low	0.08 (-0.44, 0.61)	7%	-0.18 (-0.97, 0.62)	2%
	Medium	-0.49 (-1.08, 0.11)	0%	-0.81 (-1.92, 0.32)	0%
	High	1.24 (-0.72, 3.24)	28%	2.51 (0.39, 4.67) ^{b,c}	0%
PM _{2.5} _lag1	Low	-0.46 (-1.02, 0.12)	0%	-0.03 (-0.91, 0.87)	12%
	Medium	0.84 (0.05, 1.63) ^b	47%	1.22 (0.35, 2.10) ^b	0%
	High	2.36 (0.11, 4.65) ^b	74%	3.58 (0.46, 6.81) ^b	66%
PM ₁₀ _lag1	Low	0.03 (-0.32, 0.38)	0%	0.23 (-0.43, 0.9)	8%
	Medium	0.28 (0.01, 0.55)	34%	0.25 (-0.16, 0.66)	0%
	High	0.93 (0.31, 1.55) ^b	0%	1.61 (0.73, 2.50) ^{b,c}	0%
O ₃ _lag1	Low	0.17 (-0.14, 0.49)	0%	0.44 (-0.05, 0.93)	0%
	Medium	0.24 (-0.08, 0.57)	34%	0.69 (0.07, 1.31)	55%
	High	0.67 (0.36, 0.98) ^b	0%	0.54 (0.06, 1.02)	0%

^a The 25th and 75th percentiles of daily mean temperature were used as temperature cut-offs.
^b Significantly different from the low temperature level.
^c Significantly different from the medium temperature level.

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769 **Table 3.** Pooled cumulative mortality risks (percent increase and 95% CI) of daily total
770 natural and cardiovascular mortality associated with heat exposure (99th percentile relative to
771 90th percentile of air temperature) and cold exposure (1st percentile relative to 10th percentile
772 of air temperature) by air pollutant strata.

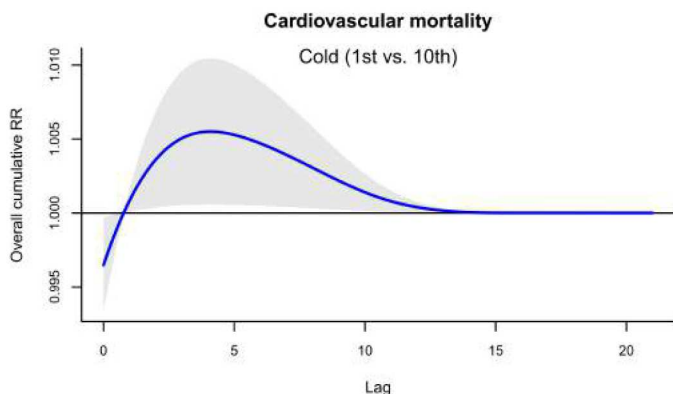
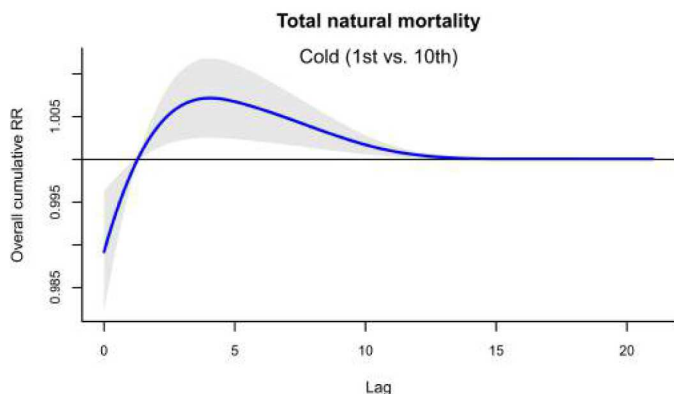
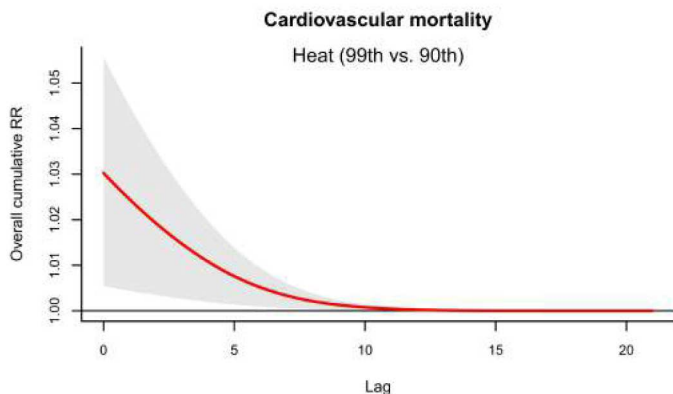
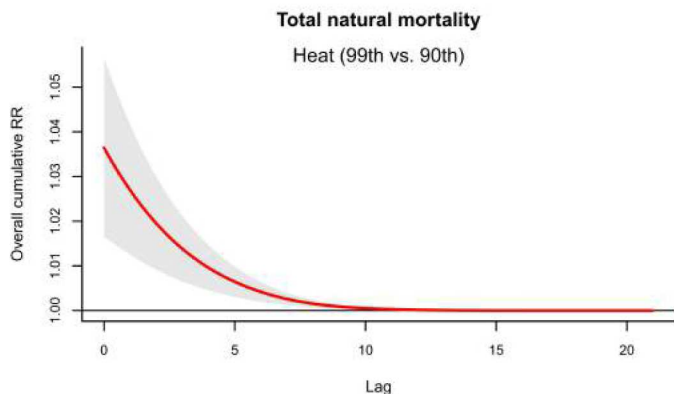
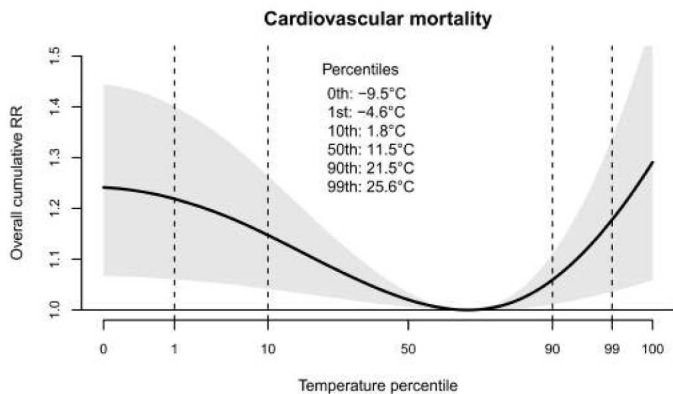
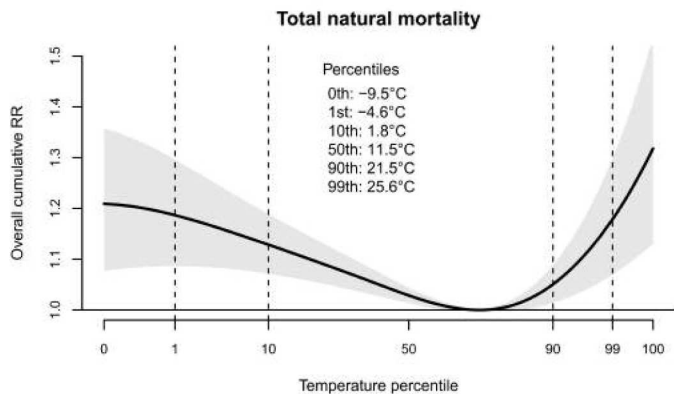
Effect	Pollutant	Pollutant levels ^a	Total natural	Cardiovascular
Heat	PNC_lag6	Low	6.94 (2.76, 11.29)	3.75 (0.29, 7.33)
		High	13.22 (-10.78, 43.67)	19.02 (-13.24, 46.68)
	PM _{2.5} _lag1	Low	4.53 (0.97, 8.21)	4.99 (-1.15, 7.56)
		High	17.71 (7.98, 28.31) ^b	16.10 (-1.62, 37.02)
	PM ₁₀ _lag1	Low	6.78 (0.53, 13.42)	7.04 (0.51, 9.69)
		High	17.39 (9.95, 25.33) ^b	13.69 (1.84, 26.91)
	O ₃ _lag1	Low	-2.08 (-4.43, 0.32)	3.90 (0.69, 7.22)
		High	14.61 (8.24, 21.36) ^b	14.83 (2.35, 28.83)
Cold	PNC_lag6	Low	3.64 (1.00, 6.35)	2.00 (0.16, 3.88)
		High	14.06 (4.45, 24.55) ^b	16.23 (3.80, 30.14) ^b
	PM _{2.5} _lag1	Low	4.00 (1.08, 7.00)	4.85 (1.71, 8.08)
		High	9.39 (-1.71, 21.74)	8.38 (-7.67, 27.21)
	PM ₁₀ _lag1	Low	4.32 (1.50, 7.21)	3.71 (0.28, 7.26)
		High	10.53 (0.24, 21.88)	14.18 (0.11, 30.22)
	O ₃ _lag1	Low	6.24 (1.72, 10.96)	6.58 (1.17, 12.29)
		High	18.39 (-31.1, 103.42)	25.75 (-51.47, 225.85)

^a The median value for each pollutant in each city was used as cut-offs for air pollution levels.

^b Significantly different from the low air pollution levels.

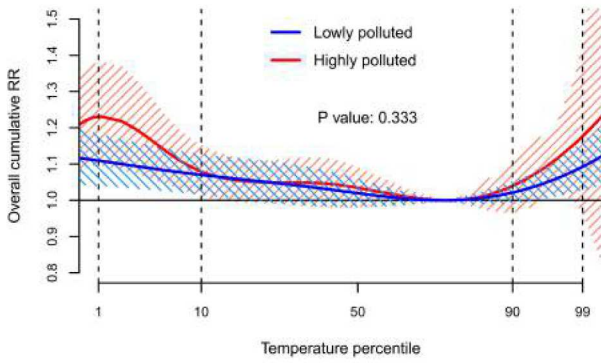
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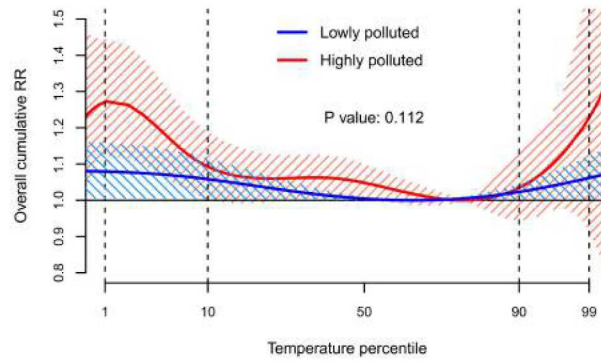


A) By PNC_lag6

Total natural mortality

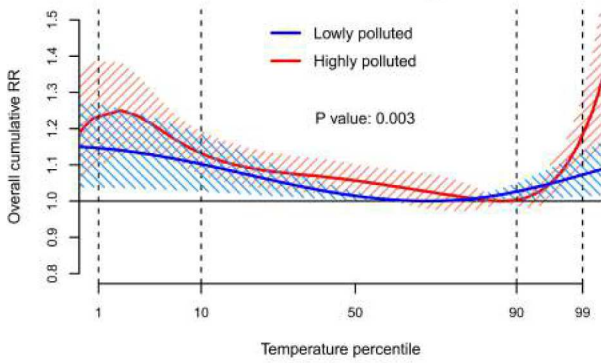


Cardiovascular mortality

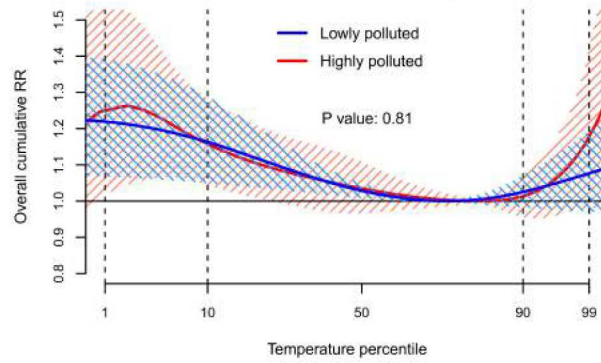


B) By PM2.5_lag1

Total natural mortality

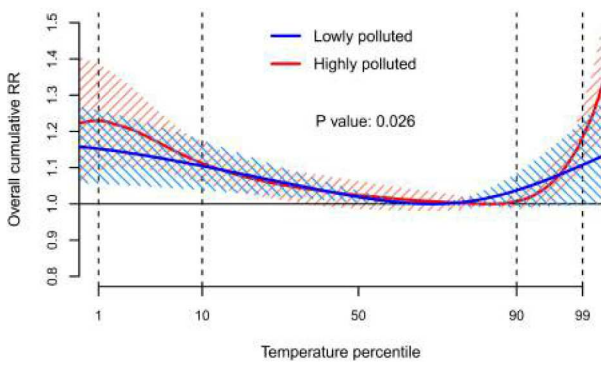


Cardiovascular mortality

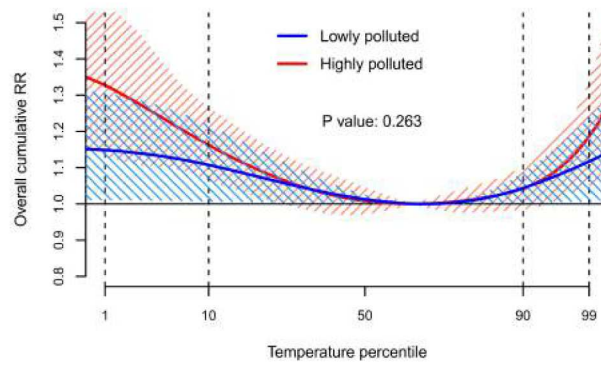


C) By PM10_lag1

Total natural mortality

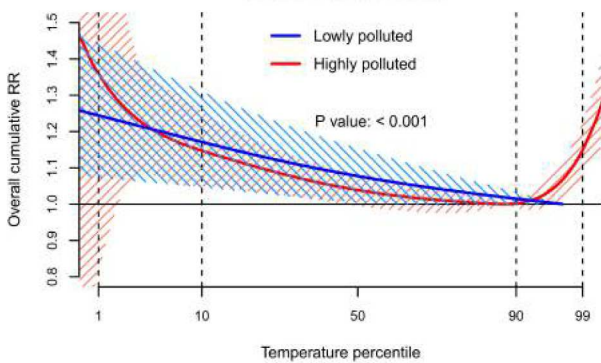


Cardiovascular mortality

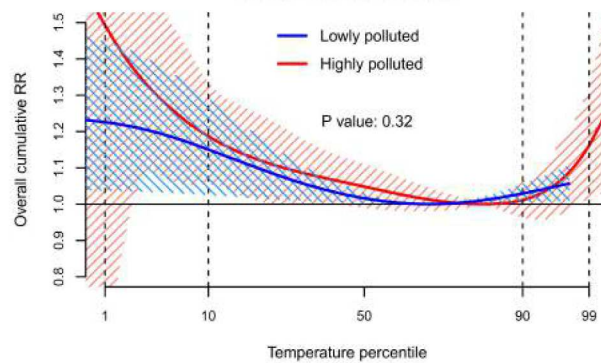


D) By O3_lag1

Total natural mortality



Cardiovascular mortality



Supplemental Information

Two-way effect modifications of air pollution and air temperature on total natural and cardiovascular mortality in eight European urban areas

Kai Chen, Kathrin Wolf, Susanne Breitner, Antonio Gasparrini, Massimo Stafoggia, Evangelia Samoli, Zorana Jovanovic Andersen, Getahun Bero-Bedada, Tom Bellander, Frauke Hennig, Bénédicte Jacquemin, Juha Pekkanen, Regina Hampel, Josef Cyrus, Annette Peters, and Alexandra Schneider, on behalf of the UF&HEALTH Study Group

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Text.S1. Description of the study areas.

Athens

In Athens, because of its high population (3.75 million in the metropolitan area) and the concentration of commercial and industrial activities in a relatively small area (approximately 450 km²), severe environmental degradation has taken place characterized by high loadings of atmospheric pollutants. The urban aerosol is strongly affected by the local air-circulation in the basin and the sea-breeze effects developing between the Attica peninsula and the surrounding sea. The wind speed increases during the day and subsides in the night.

Combustion processes used for transportation, power generation and other human activities produces a mixture of thousands chemical pollutants.

Augsburg

Augsburg is a mid-sized city with about 250,000 inhabitants in Bavaria, southern Germany. The City of Augsburg is the administrative center of the northwestern region of Bavaria. Small and middle-sized industrial entities are located in the northeastern part of the city. City planning has designed these as industrial sites. Therefore, the emissions from these sites contribute little to the air pollution concentrations in the City of Augsburg if the predominately westerly winds prevail. The terrain is moderately flat at 500 m above sea level.

Barcelona

The city of Barcelona, NE Spain, with around 1,600,000 inhabitants in the city center and over 5,000,000 in the metropolitan area, is located on the Mediterranean coast and geographically constrained by the Collserola mountain range to the west, thus atmospheric dynamics are influenced by the breeze regime. This area is characterized by mild winters, warm summers and prevalent clear sky conditions all year round. Solar radiation is thus intense and precipitations are scarce. In the urban background of Barcelona the main source of atmospheric PM is road traffic, although contributions from industry, regional secondary atmospheric pollutants, construction, and shipping are also relevant. Barcelona has one of the highest car densities in Europe (5800 cars taxed in the city/km², Barcelona City Council, 2014), with diesel vehicles making up around 47% of the fleet, and one of the most important ports of the Mediterranean. In spite of these adverse facts, PM₁₀ and PM_{2.5} mass concentrations have decreased considerably in the last decade. Moreover, different meteorological scenarios can have an impact on the levels of pollutants, such as stagnant anticyclonic conditions, recirculation of air masses, Atlantic air mass advection or African dust outbreaks.

Copenhagen

Copenhagen is a city with about 580,000 inhabitants, an area of 86 km² and a population density of 6,700 inhabitants km⁻² located on Sealand, one of the largest Danish islands. Copenhagen is the capitol of Denmark and about 1,000,000 people live in the greater Copenhagen area. There are only few industrial sites in Copenhagen and thus only little contribution from these sites is expected to affect air pollution in Copenhagen. On the other hand a harbor area is located on the eastern side of Copenhagen directly connected to the city where ships stop but also pass by the city crossing the Øresund Strait. On Amager, an island south of the inner city center the Copenhagen airport is located in a distance of less than 10 km where aircrafts take off and land with high frequency as the airport is distributing passengers to other Scandinavian distributions. The overall terrain is flat and approximately at a few meters above sea level. The climate in Copenhagen can be described as marine or continental depending on the prevailing meteorological situation. In general, westerly winds are dominating the main wind directions.

Helsinki

The city of Helsinki is the capital of Finland, with around 600,000 inhabitants. The Helsinki metropolitan area consists of four cities (Helsinki, Vantaa, Espoo, and Kauniainen) and is by far the biggest and most densely populated area of Finland with about 1 million inhabitants. Population density in certain parts of Helsinki's inner city area is very high, reaching 20,000 inhabitants per km², but as a whole Helsinki's population density of 2,741 per km² (July 2010) ranks it as quite sparsely populated in comparison to other European capital cities. Helsinki metropolitan area is located on a relatively flat land on the coast of the Gulf of Finland. The area has a humid continental climate. Average temperature varies from -5 °C in February to +18 °C in July. The vast majority of the inhabitants live in the urban areas of the cities, but within the boundaries of these cities there are also suburban and rural areas. The majority of homes are heated with district heat throughout the year, but in some areas residential wood combustion is common. The major sources of PM are traffic, energy production and residential wood combustion.

Rome

Rome is the largest Italian city, with 2.9 million inhabitants in a 1,285 km² area. The urban area is divided into five concentric circular zones, corresponding with different levels of urbanization, population density and road traffic. The historical center, which corresponds to the limited traffic zone (LTZ) (55,000 inhabitants over an area of 6 km²), shows the highest concentration of business activity in Rome. Warm months, lasting in Rome generally from

June to September, are generally characterized by large-scale high-pressure systems. Cold months (November – March) are generally characterized by moderately low temperature, prevailing north wind. Moreover periodically high pressure systems can produce temperature inversions and weak winds, leading to stagnation episodes associated with pollutant accumulation in the lower layers due to poor dispersion conditions. The main pollutant emission sources are road vehicle exhausts and small-scale combustion units, with thermal capacity < 50 MWth used in the civil sector for heating. Services and commerce are the main working activities, and emission of air pollutants from industries is relatively low compared with those originating from road traffic sources.

Ruhr Area (Essen-Mülheim-Oberhausen)

The so-called Ruhr Area is one of the most populated regions in Europe, covering 4,435 km² with about 5 million inhabitants. It has an industrial (coal mining and steel production) history that dominated the air quality for long time. While industrial emissions are still a concern, especially in certain hot-spots, traffic and domestic heating seem to dominate pollution nowadays. The three adjacent cities Essen, Mülheim and Oberhausen are located in the western part of the Ruhr Area (378.72 km²) with a total of approximately 946,000 inhabitants.

Stockholm

The city of Stockholm is the capital of Sweden, with around 900,000 inhabitants on 2014. The metropolitan area of Greater Stockholm has around 2 million inhabitants (2014). Stockholm is situated on the south-central east coast of Sweden, where the freshwater Lake Mälaren flows out into the Baltic Sea. The central parts of the city consist of fourteen islands that are continuous with the Stockholm archipelago. Over 30% of the city area is made up of waterways and another 30% is made up of parks and green spaces. The terrain is flat. Annual precipitation is around 500 – 600 mm and mean temperature around 7 °C. The main local source of PNC and PM10 is road traffic. PNC is mainly due to local vehicle exhaust, whereas PM10 is mainly from non-exhaust traffic emissions and long-range transport. Road wear is more important for PM10 concentrations in Stockholm than in many other countries where the use of studded tires is not so common. High PM10 concentrations are found in the inner city in springtime mostly due to particle suspension, as a result of the use of studded tires when roads are free from snow. Road surface wetness is very important for PM10 with systematically higher levels during dry conditions as compared to wet, while no significant difference can be observed for PNC.

Table S1. General information about the PNC measurements and measurement stations.

City/Area	Time Period	Site ID	Site type	Sampling height (m)	Monitor, sampling method, etc.	Size range
Athens	01.01.08-31.12.10	NCSR Demokritos (DEM)	Suburban background	10	TSI SMPS	10-102 nm
Augsburg	01.02.99-29.11.04	Au_KG	Urban background	2	CPC 3022A	7-3000 nm
Augsburg	30.11.04-31.12.09	Au_FH	Urban background	4	custom build TDMPS using TSI CPC 3025 and 3010	10-2000 nm
Barcelona	13.07.05-31.01.09	Ba_IJA	Urban background	10	WCPC 3785	5-1000 nm
Barcelona	09.02.09-31.12.10	BA_TG	Urban background	2	WCPC 3785	5-1000 nm
Copenhagen	15.05.01-31.12.10	Co_HCOE	Urban background	20	custom build DMPS using TSI CPC 3010	10-110 nm
Helsinki	06.01.01-01.10.04	Kumpula1	Urban background	18	DMA TSI307 + CPC TSI3022	10-100 nm
Helsinki	13.10.04-31.12.10	Kumpula SMEAR III	Urban background	4	Hauke-type DMA + CPC TSI3025 Hauke-type DMA + CPC TSI3010	10-100 nm
Rome	12.04.01-31.12.10	Rome_15	Urban traffic	2	TSI 3022A	7-3000 nm
Ruhr Area	01.03.09-31.12.13	E_MH_OB	Urban background	4	TSI SMPS 3936/DMA 3080/UWCPC 3786	14-750 nm
Stockholm	01.05.01-31.07.05	Sto_Rosenlundsgatan	Urban background	20	TSI 3022	7-3000 nm
Stockholm	17.02.08-31.12.10	Sto_Torkel	Urban background	24	TSI 3755	4-3000 nm

Table S2. Percentage of missing data for daily deaths, air pollution, and air temperature in eight European cities

	Helsinki	Stockholm	Copenhagen	Ruhr Area	Augsburg	Rome	Barcelona	Athens
<i>Daily mortality</i>								
Nonaccidental	0	0	0	0	0	0	0	0
Cardiovascular	0	0	0	0	0	0	0	0
<i>Air pollutants</i>								
PM _{2.5} (µg/m ³)	0.2	0.0	31.8	51.0	3.2	50.1	6.2	28.4
PM ₁₀ (µg/m ³)	0.1	0.3	28.0	1.8	0.8	0.1	6.2	5.2
O ₃ (µg/m ³)	0.0	1.7	28.2	8.4	0.7	0.3	2.3	16.4
PNC (particles/cm ³)	3.6	23.1	48.1	16.5	16.7	39.1	33.2	45
<i>Air temperature (°C)</i>								
	0	0	3.3	3.7	2.0	0.6	1.1	0

Table S3. Average air temperature distribution (°C) at different air pollution levels in eight European cities. ^a

Air pollution levels	Min	1st percentile	10th percentile	25th percentile	50th percentile	75th percentile	90th percentile	99th percentile	Max
PNC levels									
Low	-9.2	-4.2	2.6	7.1	13.1	18.5	21.6	25.5	28.1
High	-8.3	-4.6	1.2	5.5	10.2	15.5	19.9	25.0	27.5
PM _{2.5} levels									
Low	-7.3	-3.1	2.5	6.4	11.6	16.9	20.4	24.3	27.3
High	-8.6	-4.9	1.5	5.6	11.2	18.3	22.3	26.0	28.1
PM ₁₀ levels									
Low	-8.5	-3.7	2.0	5.9	11.5	17.4	20.7	24.5	27.3
High	-8.9	-4.1	2.0	6.2	11.7	18.3	22.2	26.1	28.4
O ₃ levels									
Low	-9.5	-5.6	0.0	3.8	7.6	12.6	16.6	21.2	24.4
High	-6.4	-0.9	5.8	10.8	16.5	20.5	23.1	26.5	28.4

^aThe median value for each pollutant in each city was used as cut-offs for air pollution levels.



Fig.S1. Locations of the eight European urban areas in the UF&HEALTH study.



Fig.S2. Pearson correlation matrix plots between PNC, PM_{2.5}, PM₁₀, O₃, and air temperature in eight study areas.

PNC_lag6 and daily mortality

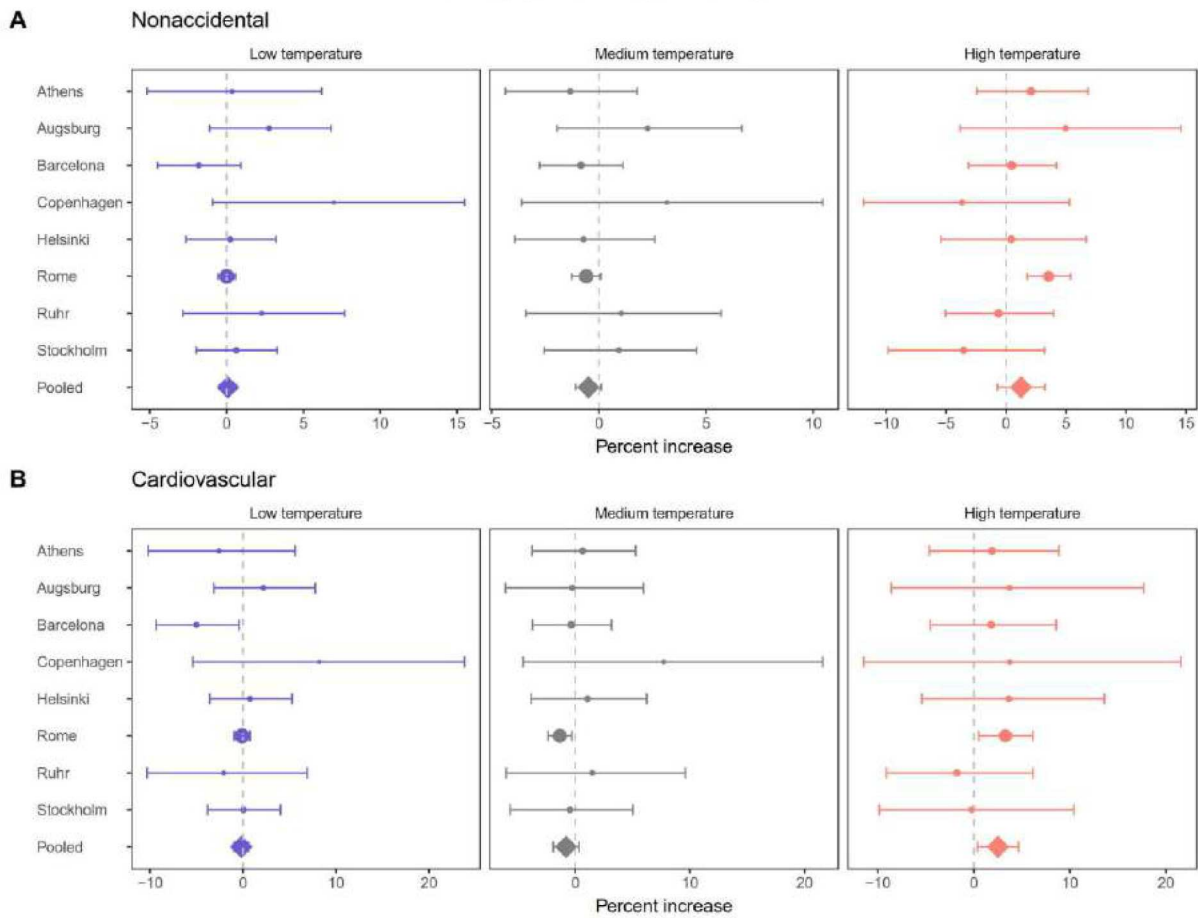


Fig.S3. City-specific and pooled estimates on the short-term association between PNC and daily cause-specific mortality: percent increase (95% CI) per 10,000 particles/cm³ increase in PNC at lag 6. (A) Nonaccidental mortality; (B) cardiovascular mortality. The point size for city-specific estimates represent the weights used in the random effects meta-analyses. The diamonds represent the pooled estimates.

PM2.5_lag1 and daily mortality

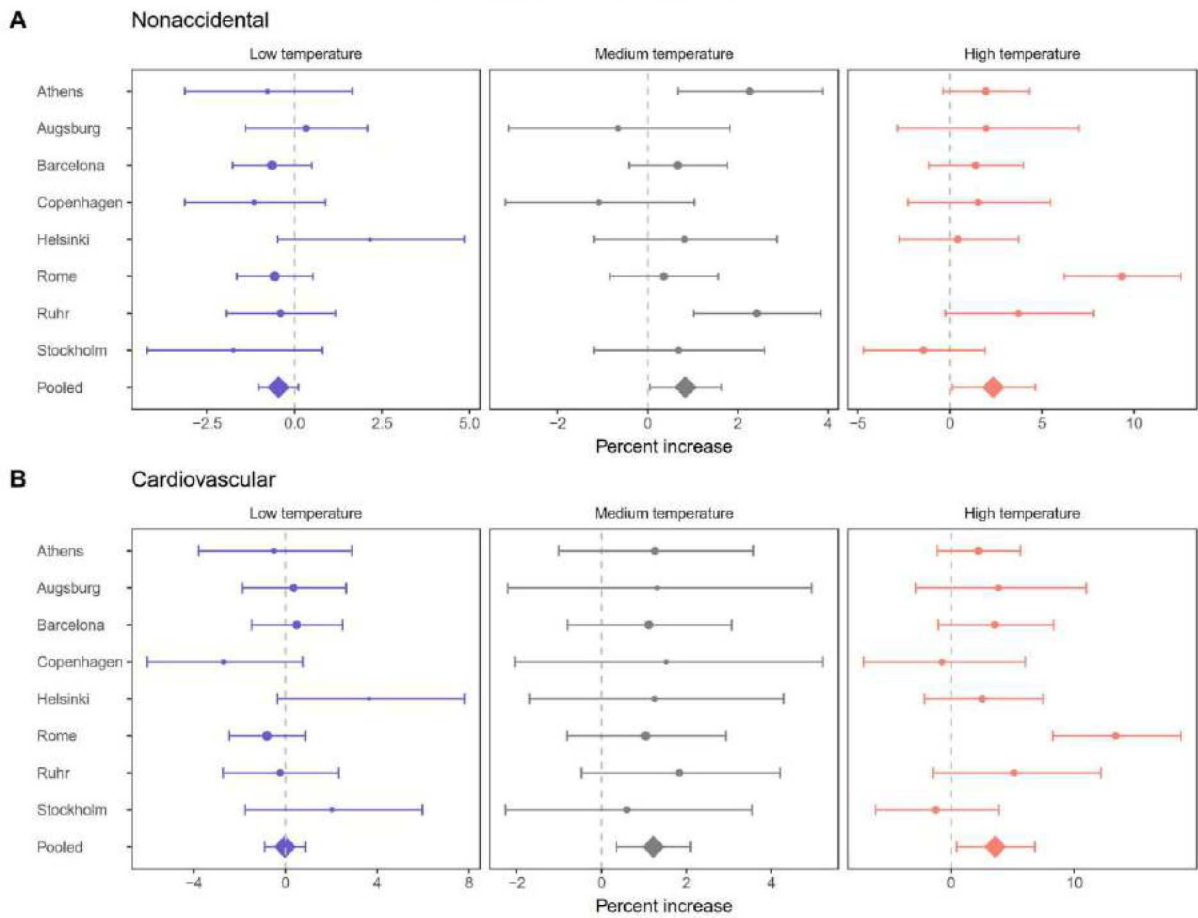


Fig.S4. City-specific and pooled estimates on the short-term association between PM_{2.5} and daily cause-specific mortality: percent increase (95% CI) per 10 µg/m³ increase in PM_{2.5} at lag 1. (A) Nonaccidental mortality; (B) cardiovascular mortality. The point size for city-specific estimates represent the weights used in the random effects meta-analyses. The diamonds represent the pooled estimates.

PM10_lag1 and daily mortality

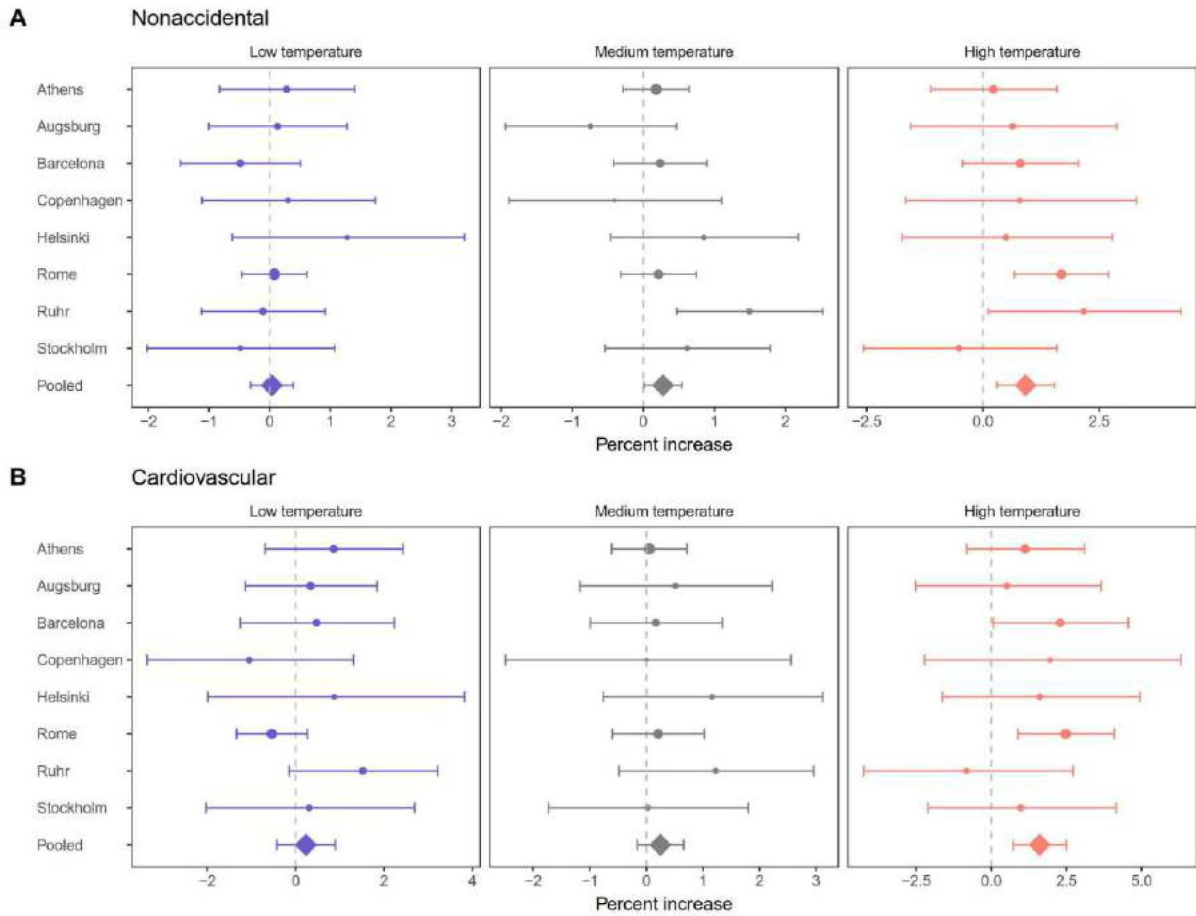


Fig.S5. City-specific and pooled estimates on the short-term association between PM₁₀ and daily cause-specific mortality: percent increase (95% CI) per 10 $\mu\text{g}/\text{m}^3$ increase in PM₁₀ at lag 1. (A) Nonaccidental mortality; (B) cardiovascular mortality. The point size for city-specific estimates represent the weights used in the random effects meta-analyses. The diamonds represent the pooled estimates.

O3_lag1 and daily mortality

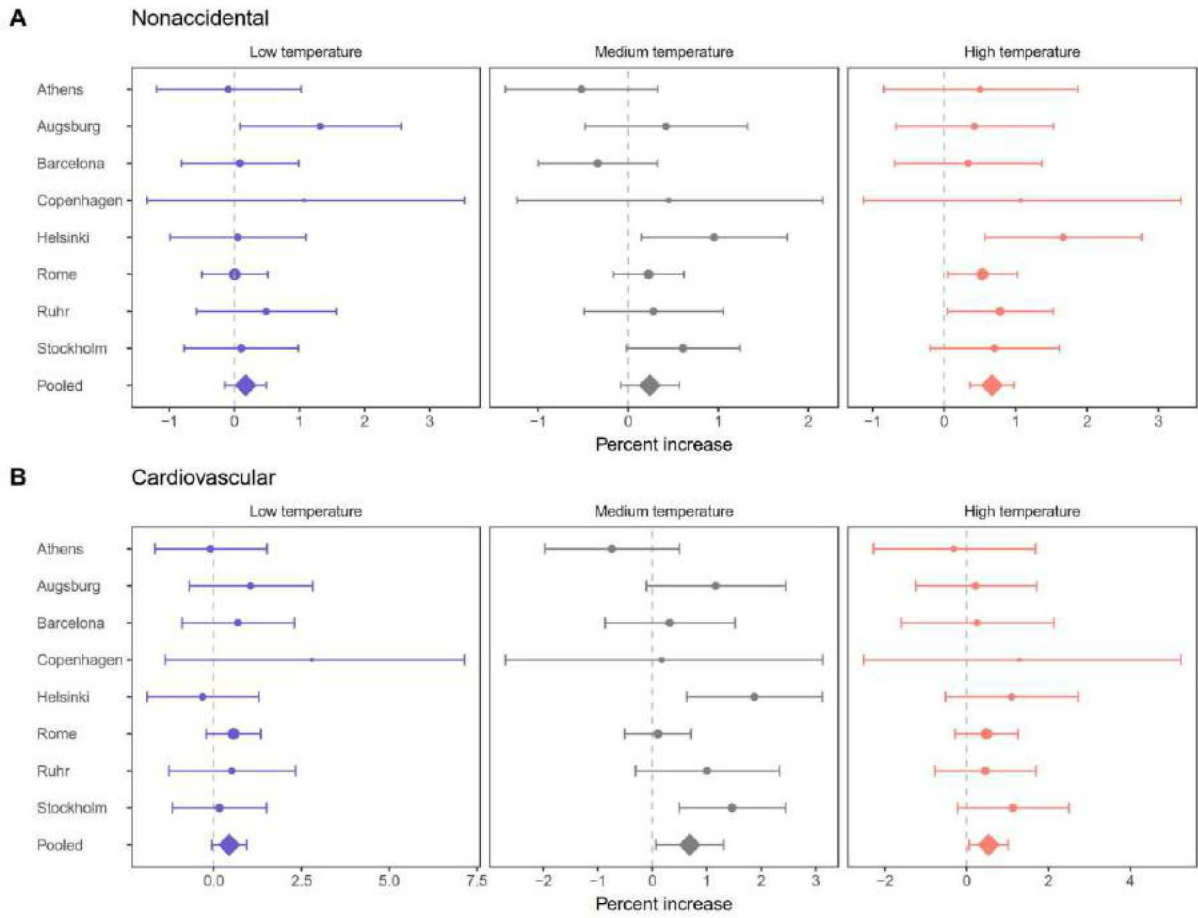


Fig.S6. City-specific and pooled estimates on the short-term association between O₃ and daily cause-specific mortality: percent increase (95% CI) per 10 µg/m³ increase in O₃ at lag 1. (A) Nonaccidental mortality; (B) cardiovascular mortality. The point size for city-specific estimates represent the weights used in the random effects meta-analyses. The diamonds represent the pooled estimates.

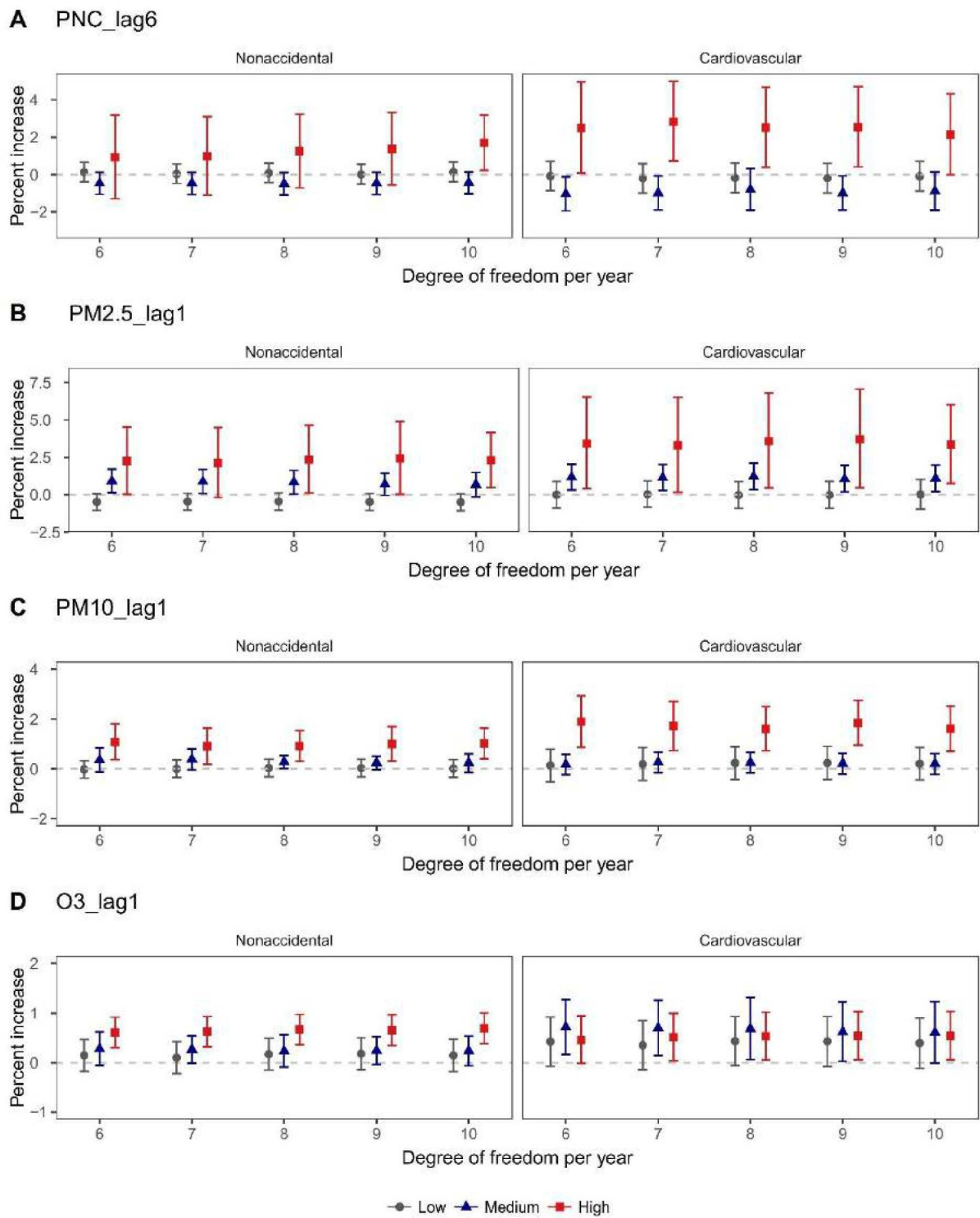


Fig.S7. Sensitivity analysis of modified air pollution-related mortality risk estimates (95% CI) by temperature levels (low, medium, high) using 6-10 degree of freedom (df) per year for time trend.

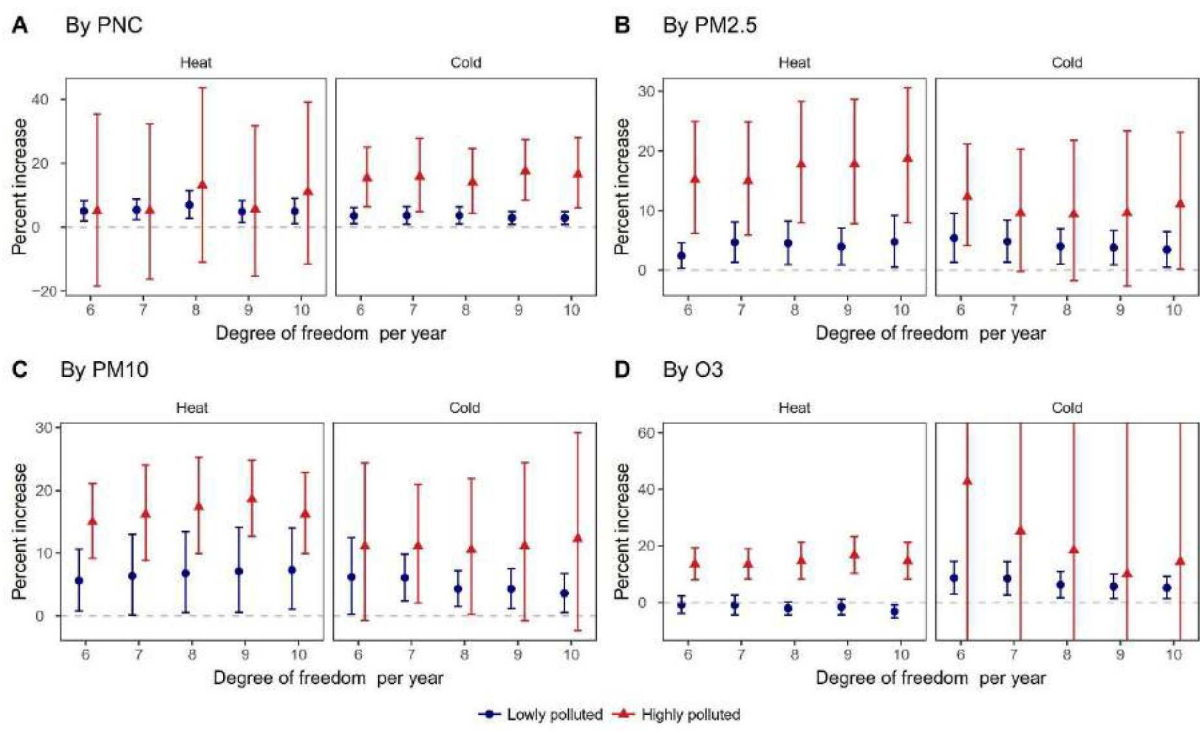


Fig.S8. Sensitivity analysis of modified heat- and cold-related nonaccidental mortality risk estimates (95% CI) by air pollution strata using 6-10 degree of freedom (df) per year for time trend. For cardiovascular mortality, results were similar.

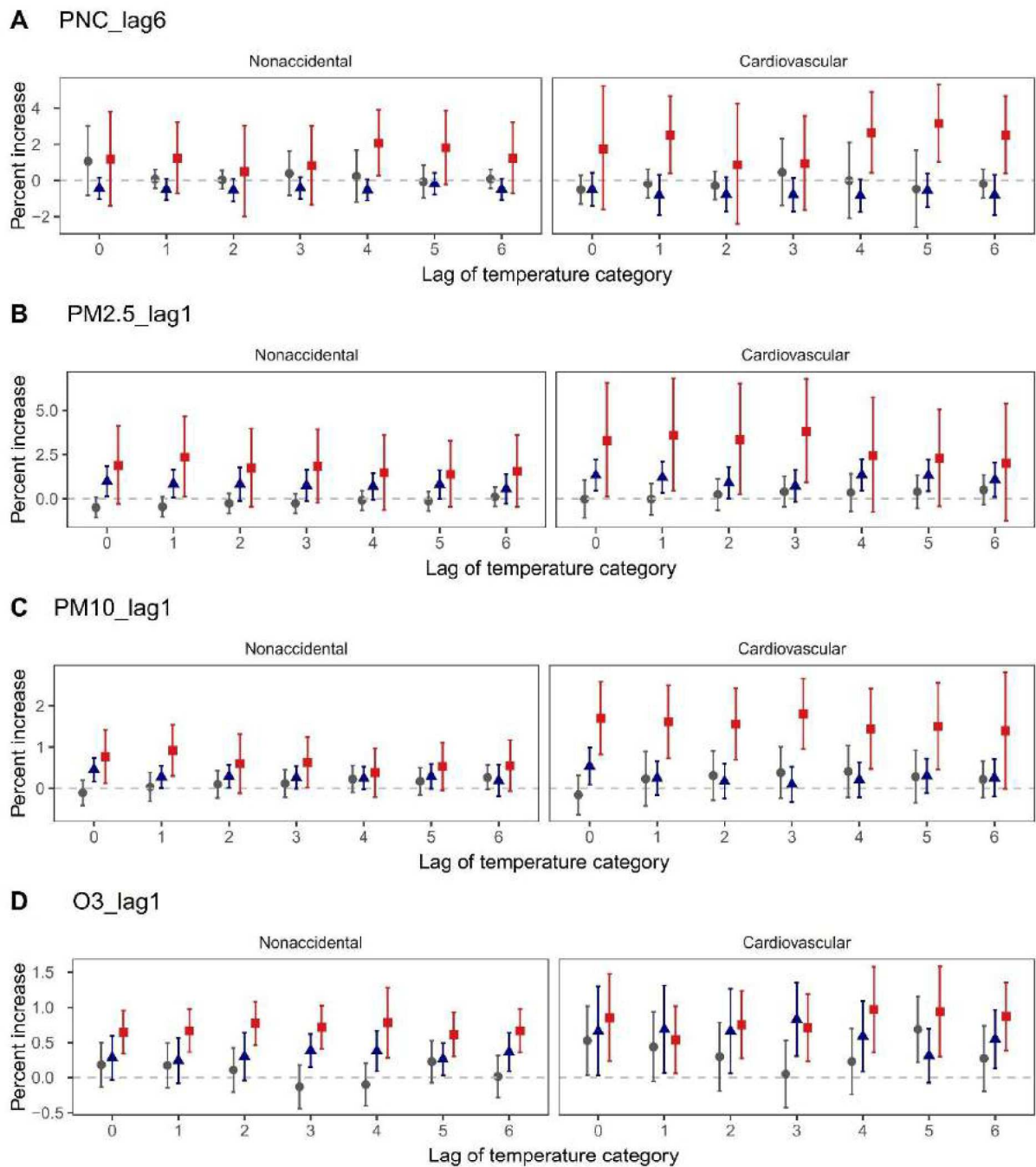
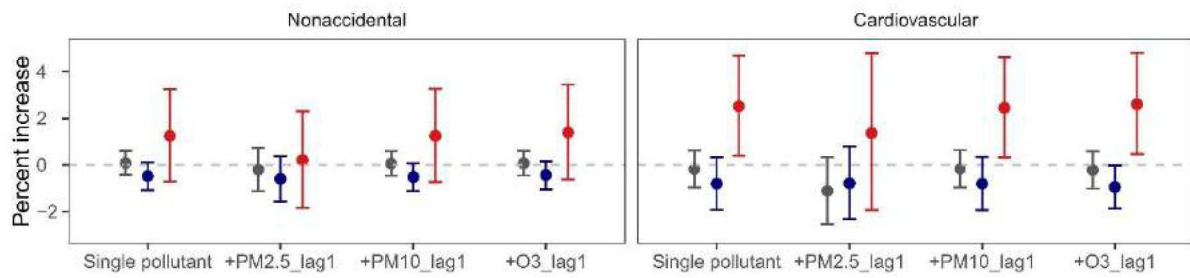
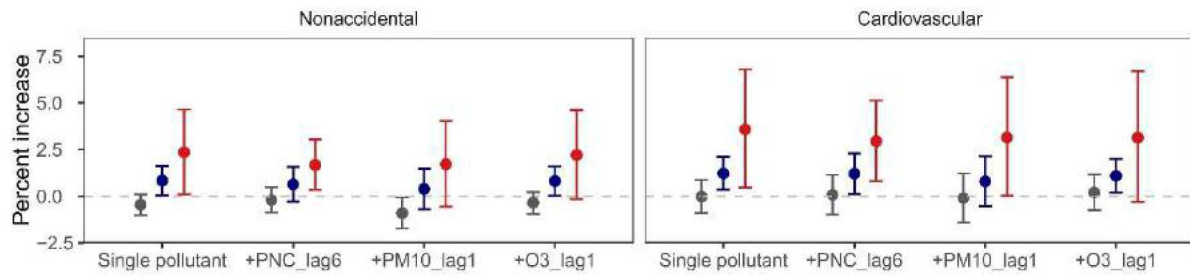
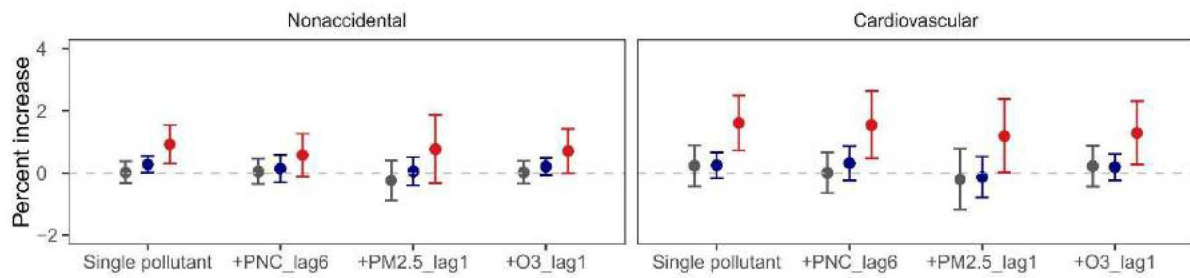
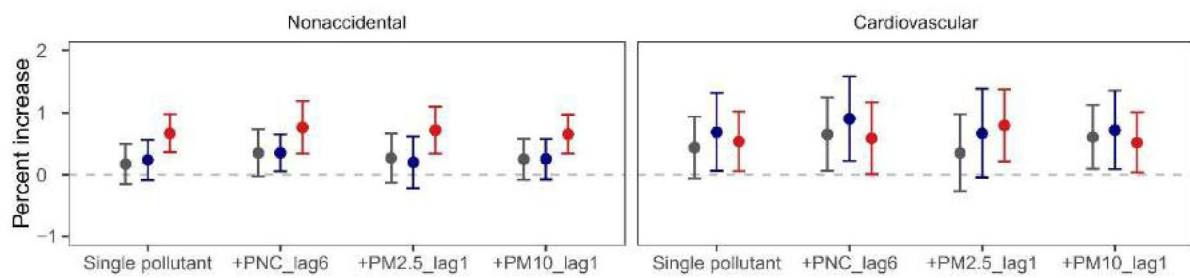


Fig.S9. Sensitivity analysis of modified air pollution-related mortality risk estimates (95% CI) by air temperature levels (low, medium, and high) using different lag days for air temperature category in eight European urban areas. Risk estimates are presented as percent increase (95% CI) in daily cause-specific mortality associated with a 10,000 particles/cm³ increase in PNC at lag day 6 or a 10 µg/m³ increase in PM_{2.5}, PM₁₀, and O₃ at lag day 1.

A PNC_lag6**B PM2.5_lag1****C PM10_lag1****D O3_lag1**

● Low ● Medium ● High

Fig.S10. Modified air pollution-related mortality risk estimates (95% CI) by air temperature levels (low, medium, and high) with and without adjustment for co-pollutants in eight European urban areas. Risk estimates are presented as percent increase (95% CI) in daily cause-specific mortality associated with a 10,000 particles/cm³ increase in PNC at lag day 6 or a 10 µg/m³ increase in PM_{2.5}, PM₁₀, and O₃ at lag day 1.

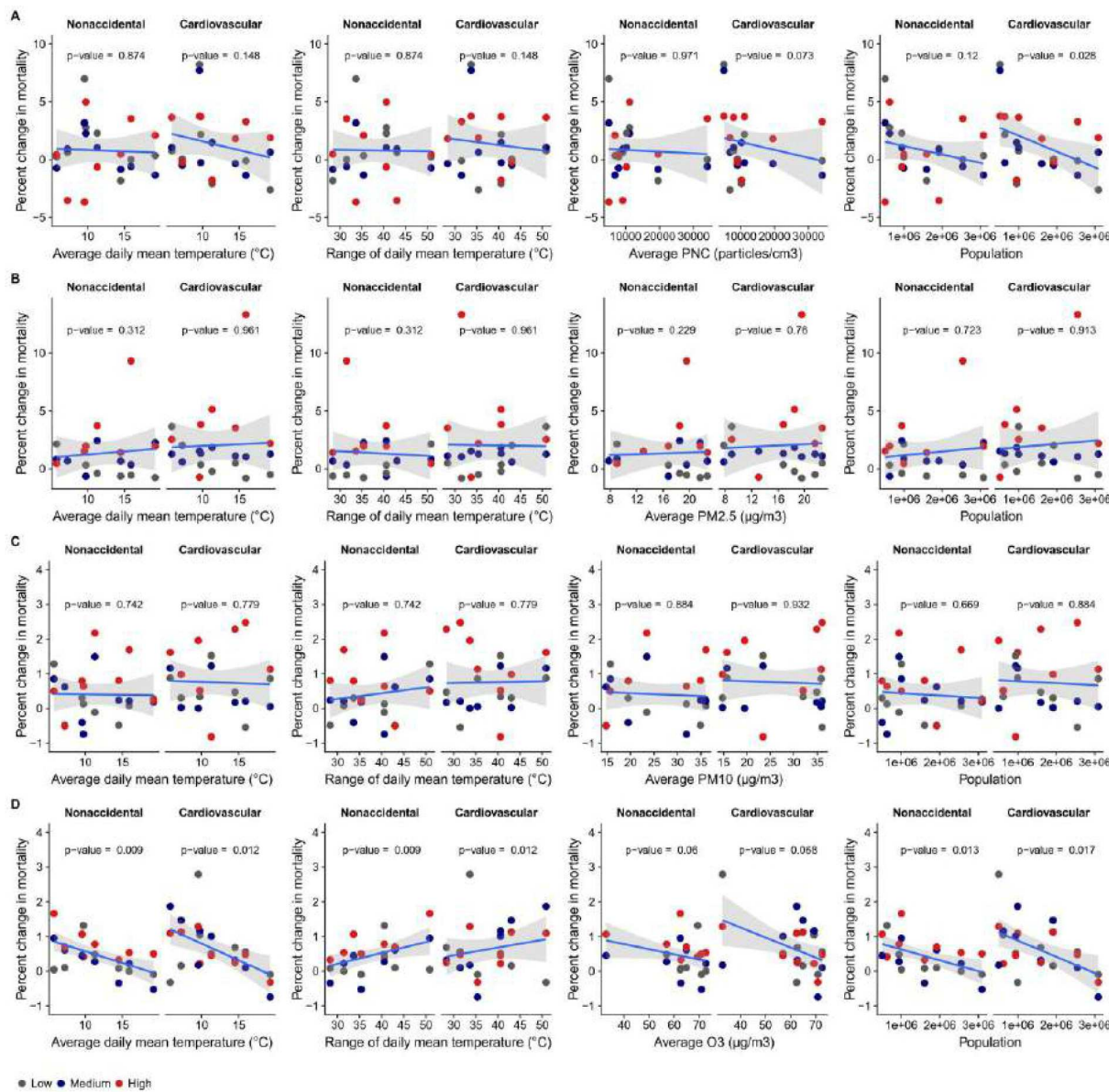


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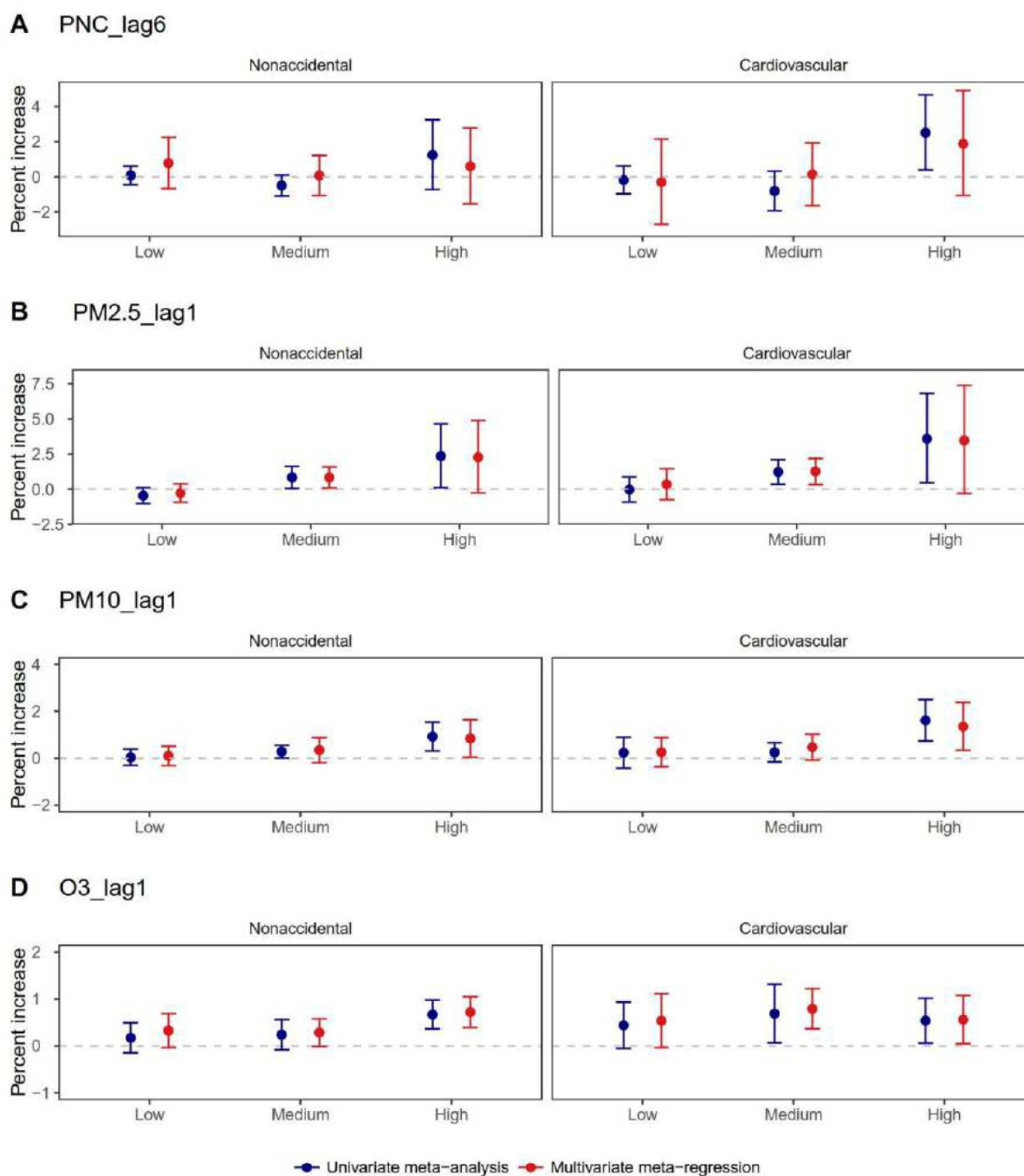


Fig.S12. Sensitivity analysis of modified air pollution-related mortality risk estimates (95% CI) by air temperature levels (low, medium, high) using multivariate meta-regression models, which included city-specific average temperature, temperature range, and population as additional meta-predictors.

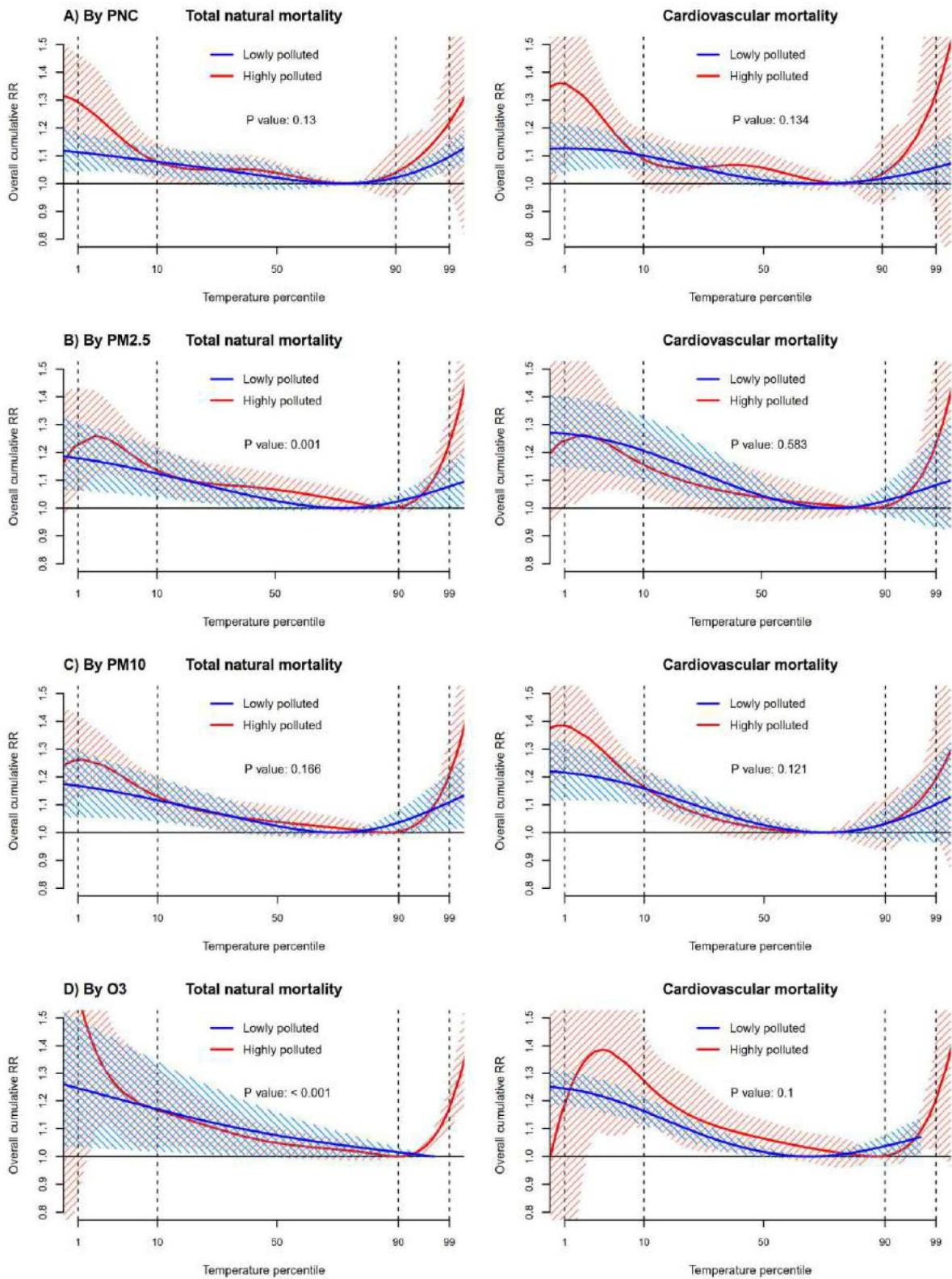
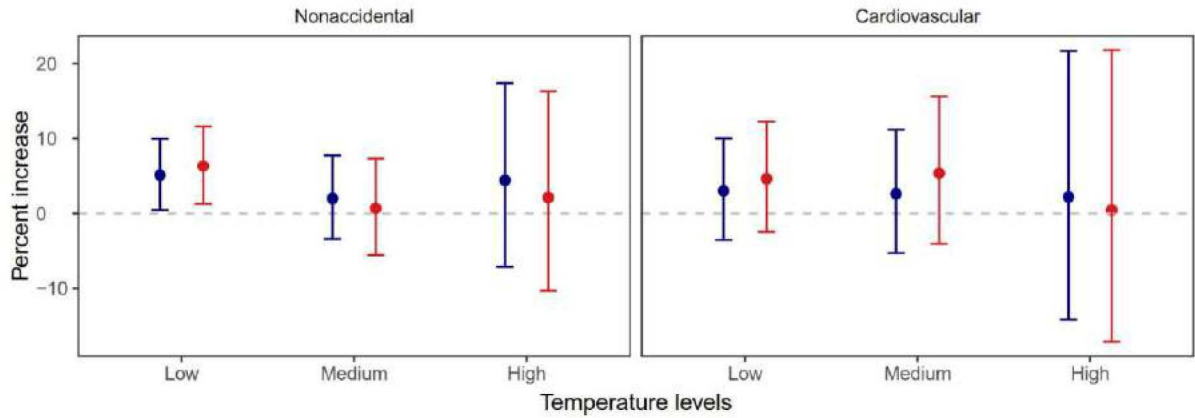


Fig.S13. Sensitivity analysis of modified overall cumulative air temperature-mortality associations by air pollution with 95% CIs using multivariate meta-regression models, which included city-specific average temperature, temperature range, and population as additional meta-predictors.

A Effect modification by temperature



B Effect modification by air pollution

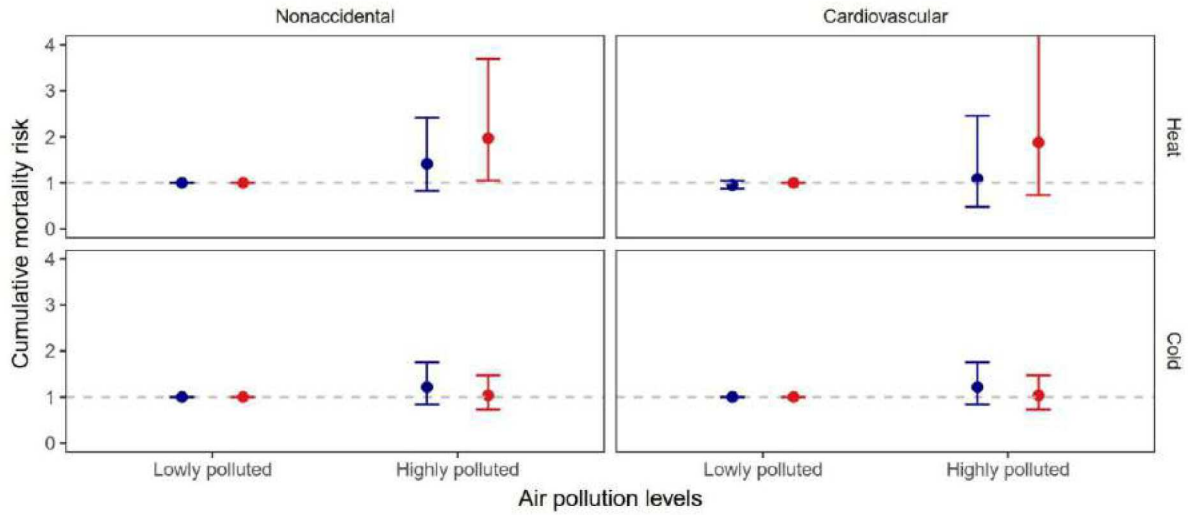


Fig.S14. Sensitivity analysis of using UFP and total PNC in Augsburg, during 2004-2009.