

Analyzing effect modifiers of the temperature-mortality relationship in the Paris Region to identify social and environmental levers for more effective adaptation to heat

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- 2 identify social and environmental levers for more effective adaptation to heat
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23

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29	
30	Abstract
31	Adaptation to heat is a major challenge for the Paris region (France). Based on fine-scale data
32	for the 1,287 municipalities of the region over 2000-2017, we analyzed (time-series design)
33	the temperature-mortality relationship by territories (urban, suburban, rural), age (15-64 and
34	≥65) and sex, and explored how it was modified by vegetation and socio-economic indicators.
35	Heat was associated with an increased mortality risk for all territories, age groups, sex, and
36	mortality causes. Women aged 65 and over residing in the most deprived municipalities had a
37	relative risk (RR) of death at 29.4°C (compared to 16.6°C) of 4.2 [3.8:4.5], while the RR was
38	3.4 [3.2:3.7] for women living in less deprived municipalities.
39	Actions to reduce such sex and social inequities should be central in heat adaptation policy.
40	Keywords: mortality, temperature, heat, climate, sex inequities, social inequities, vegetation
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60 1. Introduction

61 Heat and heat extremes are more than ever a public health priority in Western Europe (IPCC 2022, 62 Rousi, Kornhuber et al. 2022). In France, the intensity of recent heat waves has been fueled by 63 climate change. For example, the likelihood of observing the extreme temperatures recorded in 2019 64 (World weather attribution 2019) or in 2022 (Météo-France 2023) in France would have been very low without human influence on the climate. Between 2014 and 2022, more than 32,000 excess deaths in 65 66 France were due to heat, 28% of which occurred during heat waves (Pascal, Wagner et al. 2023). The 67 largest impacts were observed in the warmest years: 2022 (6,969 [6,277:7,445] deaths attributable to 68 heat) and 2019 (4,441 [4,086:4,717] deaths attributable to heat).

69 Heat is of particular concern in the Paris region, where more than 12 million inhabitants are exposed to 70 rapidly changing temperatures. Urban areas cover 27% of the region, agricultural land (mainly cereals) 71 49%, and natural areas (forests, water) 24%. The city of Paris is one of the most densely populated 72 cities in Europe, with more than 20,600 inhabitants per square kilometer on average, and up to 41,500 73 inhabitants per square kilometer in some neighborhoods. The urban heat island (UHI) effect can reach 74 up to 6 to 8°C between the moist urbanized and central areas and the outskirts of the region 75 (Lemonsu and Masson 2002, Orlando, Mosconi et al. 2021) (Madelin and Dupuis 2020). The Paris 76 region is also characterized by large economic inequities (Sagot 2019).

77 In the city of Paris, the fraction of mortality due to extreme heat has increased since the 2000s 78 (Pascal, Wagner et al. 2023). However, few studies have examined the individual, social and 79 environmental risk factors associated with heat-related health risks in the whole Paris region. A case-80 control study focusing on the 2003 heat wave and restricted to urban areas reported an increased risk 81 for people with pre-existing health conditions (e.g. mental disorders, neurological or cardiovascular 82 diseases, cancer...), for manual workers, for people living in poorly insulated buildings, and for people 83 with a bedroom under the roof (Vandentorren, Bretin et al. 2006). The average minimum temperature 84 observed during the days before deaths, used as a proxy of the UHI, was also associated with an 85 increased mortality risk in 2003 (Laaidi, Zeghnoun et al. 2012). An ecological study over the period 86 1990-2015 and covering the whole region found that the risk of heat-related mortality was higher in 87 municipalities that lacked green spaces or had a high proportion of impervious surfaces (Pascal, Goria 88 et al. 2021).

89 In terms of prevention, current adaptation to heat is mainly based on communication of appropriate 90 behavior all-through summer (e.g. how to stay hydrated, how to reduce indoor temperatures, how to 91 recognize heat stroke), and on short-term measures triggered by heat warnings (e.g. increasing 92 staffing in hospitals, cancelling sports events, shifting working hours). People who feel vulnerable to 93 heat can also contact the municipal authorities to benefit from at-home visits during warnings. In 2022, 94 of the2.1 million inhabitants of the city of Paris, only 10,000 people registered themselves as 95 vulnerable to heat (Ville de Paris 2023). Several municipalities of the region have also launched 96 programs to increase vegetation and shade in public spaces. However, long-term adaptation 97 strategies to simultaneously reduce the three dimensions of heat vulnerability (exposure, individual 98 susceptibility and adaptive capacity), proposed by Ellena, Breil et. al. (Ellena, Breil et al. 2020), are still 99 needed. The nexus between heat, vegetation, socio-economic status and health is of particular 100 interest in this context. Figure 1 schematizes the complex interactions between those factors. Through 101 shadow and evapotranspiration, vegetation has a direct influence on outdoor temperature (Shashua-102 Bar, Pearlmutter et al. 2009), which in turn affects vegetation (Kabano, Lindley et al. 2021). Outdoor 103 temperature is classically used as a proxy of indoor and outdoor temperature exposure in 104 epidemiological studies. It is consistently associated with mortality. Actual exposure depends on 105 individual behaviors, on health conditions (for instance affecting mobility or behavioral capacities), on 106 socio-economic status (for instance through occupational exposure, or access to air conditioning), and 107 on vegetation (for instance, providing access to fresh areas). Vegetation is also beneficial to social 108 link, physical and mental health, reducing the overall vulnerability of the population to environmental 109 hazards such as heat (Yang, Zhao et al. 2021). Those interdependencies may additionally depend on 110 the type of territory, from rural to urban, but also on gender and age.





113 and health

114 To identify levers for adaptation, we investigated the temperature-mortality relationship by type of 115 territory, age and sex, and explored how it was modified by vegetation and socio-economic indicators.

116 2. Method

117 2.1. Study area and study period

118 The region is divided into eight administrative departments comprising 1,287 municipalities (using the 119 2019 geographical reference of the National Institute for Statistics and Economic Studies (Insee)). The 120 most central departments (Paris, Hauts de Seine, Seine-Saint-Denis and Val-de-Marne) are also the 121 most urbanised and densely populated. The other departments (Val-d'Oise, Yvelines, Essonne, Seine-122 et-Marne) are less densely populated and more rural. The regional planning authority (Institut Paris 123 Region) further classifies the departments and municipalities into six territorial sub-divisions, from rural 124 areas to the densest part of Paris. This territorial division based on land use is more interesting for 125 studying the influence of heat on health than other administrative divisions. In order to ensure sufficient statistical power for the epidemiological analyses, these six territorial sub-divisions were 126 127 merged into three larger territories representing 1) rural, small and medium-sized municipalities, 2) the 128 Paris agglomeration, 3) the inner metropolitan area and the hyper-centre (Figure 2).

Analyses focused on the summer months (April-September) of the period 2000-2017. More recent
years could not be included in the analyses due to unavailability of mortality data by causes.

Territorial division of the Region of Paris, 2022 based on population density, urban attractivity and building continuity



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132 Figure 2 – Territorial sub-division of the Paris region used to stratify the analysis

133 2.2. Mortality and population data

Daily death counts for all-causes, cardiovascular (ICD-10 I00-I99), cardiac (I00-I52) and respiratory (J00-J99) causes were collected for each municipality (CépiDC data), by sex and age (15-64 years and >=65 years).

137 Cardiovascular and respiratory mortality are the most frequently investigated causes in the literature 138 (Cheng, Xu et al. 2019). Cardiovascular mortality is a broad category of diseases, engaging different 139 physiological systems in response to heat. We decided to include also cardiac causes to capture 140 possible differences for this subset of cardiovascular diseases. Due to the small number of cases, it 141 was not possible to include other categories of cardiovascular mortality such as cerebrovascular 142 mortality for instance.

For each municipality, annual total and 200-m gridded population data were provided by Insee for the 144 1999 and 2006-2017 censuses. Linear interpolations were used to estimate the population for years 145 between censuses.

146 2.3. Temperature data

Daily gridded minimum (Tmin), maximum (Tmax) and mean (Tmean) temperatures were produced by spatially interpolating temperatures recorded by Météo-France weather stations. The interpolation took into account the influence of altitude and urbanisation (Kounkou-Arnaud and Brion 2018, Le Roy, Lemonsu et al. 2020) (Lemonsu, Alessandrini et al. 2024). The horizontal resolution for these griddedtemperatures was 1.25 km across the whole region.

Two additional indicators derived from microclimatic modelling carried out over the same period 2000-2017 (with a horizontal resolution of 1 km) were considered: the daily mean radiant temperature (MRT) and the daily mean universal thermal climate index (UTCI) (method described in Supplementary material).

For each municipality, daily temperature indicators were estimated by combining the griddedtemperature at a 1.25 or 1 km resolution with the 200-gridded population data. The values for each temperature grid intersecting the surface of the municipality were weighted and averaged according to the population at 200 m. (Forceville, Lemonsu et al. 2024). This approach provides an estimate of the average exposure in each municipality, giving more importance to inhabited areas.

161 Supplementary Figure S1 illustrates the geographical variations of each indicator for a selected day 162 (the 7th of July 2015).

163 2.3. Vegetation data

Three complementary indicators were considered to assess the level of greening of each municipality and investigate if this modified the temperature-mortality relationship. Here, the objective was not to capture the daily influence of vegetation on temperature (this being already captured in the fine-scale temperature data), but to investigate the broader possible influences of vegetation.

The first indicator, based on land-use, estimated the proportion of artificialized areas within each municipality. It was calculated for 2006, 2012 and 2018, as the proportion of the municipal area classified in items 11 to 13 (11.Urban fabric, 12.Commercial and transport units, 13. Mines, landfills and construction sites) in the Corine Land Cover database (European Copernicus programme), as described in Pascal et. al (Pascal, Goria et al. 2021). It indicates the possible presence of permeable soil and vegetation. The second indicator was the annual mean summer (June-August) Normalised Difference Vegetation Index (NDVI). It was estimated for each municipality as a surface-weighted average of Landsat data on a 30-metre resolution grid (Hough, Rolland et al. 2023). It indicates the presence of vegetation without information on the type of vegetation.

The third indicator was the proportion of the municipality covered by tree canopy. It was computed as a surface-weighted average of the density of the urban canopy estimated for 2015 (between March-October) from Sentinel 2, Landsat 8 data on a 20m resolution grid (Copernicus 2015). It indicates the presence of trees.

182 Maps of each vegetation indicators for the year 2015 are presented on Supplementary Figure S2.

183 2.4. Socio-economic data

For each municipality, the median income (in \in), % of high school graduates, % of unemployment, % 184 185 of manual workers and % of social housing were obtained from Insee. The median income, % of high 186 school graduates, % of unemployment, % of manual workers were used to calculate the French Depri-187 vation Index (FDep) according to the method described by Rey et al. (Rey, Jougla et al. 2009). The 188 FDep is a composite indicator of social disadvantage that was specifically designed for use in health 189 studies in France. We also considered the median income and the percentage of social housing as 190 two additional indicators of social inequities. The median income exhibits larger variations in the Paris 191 Region than in other regions of France. Housing is also a strong driver of social inequities in the region 192 (Sagot 2019). Social housing rents are regulated and, on average, two to three times lower than regu-193 lar rents (Ministère de la transition écologique et de la cohésion des territoires 2023). The majority of 194 social housing residents are from low-income households.

195 Maps of each socio-economic indicators for the year 2015 are presented on Supplementary Figure S3.

196 **2.5. Statistical analysis**

197 2.5.1. General approach

Time-series analyses using non-linear distributed lag models (DLNMs) were used to consider the nonlinear relationship between temperature and mortality, as well as the delayed effects of temperature (Gasparrini, Armstrong et al. 2010, Gasparrini, Scheipl et al. 2017, Gasparrini 2022, Song, Gasparrini et al. 2023). 202 Daily mortality rates at the municipality level were modelled using a quasi-Poisson regression to 203 account for over-dispersion:

204 $log[E(Y_{ij})] = \alpha + \beta_1 dow_j + \beta_2 bh_j + ns(time_j) + ns(day_j)^*year + f(Tmean_{ij},l) + \beta_3 prop75_i + \beta_4 FDep_i + \beta_5$ 205 $NDVI_j$

where Y_{ij} is the number of deaths in small area i in the day j. The annual population size of each municipality was included as an offset to account for the difference in population size between municipalities. Controlling factors such as FDep (*FDep*_i), the percentage of people aged 75 and over (*prop75*_i), NDVI, day of the week (*dow*_j), bank holidays (*bh*_j), seasonality, and long-term trends were also included in the models. The function f(Tmean_{ij},I) specifies the cross-basis that describes the association of mortality and daily mean temperature *Tmean*_{ij} and its lag structure and is characterised by two dimensions:

-the exposure-response dimension, modelled using a quadratic B-spline with two internal
nodes located at the 50th and 95th percentiles of the temperature indicator distribution, based
on the Akaike criteria;

-the lagged-response dimension, fitted using a natural spline with two internal nodes, placed at
equally spaced values in the log-scale to allow for more flexibility in the first lags where more
variability is expected. Individual lags were investigated up to 10 days.

To capture seasonality, a natural spline of day of the season, defined as a count in days increasing from April to September, with 3 degrees of freedom per summer was used (ns(day)). An interaction between this spline function and the year indicator was specified to allow the seasonal trend to change across the years. The long-term mortality trend was controlled by a natural spline of time with one degree of freedom every 10 years (ns(time)).

A sensitivity analysis included a spatial component in the model to account for residual spatial autocorrelation and heterogeneity. It was modeled with a Markov random field smoother using a conditional autoregressive model with a normally distributed conditional autoregressive prior to deal with the neighbourhood correlations according to whether municipalities shared a common boundary (Chien, Guo et al. 2016, Pascal, Goria et al. 2021). This modelling philosophy was used to develop several models summarized in supplementary material Figure S4, and detailed below. To ensure sufficient statistical power, those models were developed only when at least 5,000 deaths were recorded during the study period, as proposed by Masselot et al. (Masselot, Mistry et al. 2023).

All analyses were performed using R software with the *dlnm* and *mgcv* packages (Gasparrini 2011,

Wood 2017), and the script developed for small-scale analysis (Gasparrini 2022).

235 2.5.2. Selection of the temperature indicator

Initially, models for all-causes, cardiovascular, cardiac and respiratory mortality were built for the entire region, using various temperature indicators. Those models were stratified by age-group, under the assumption that the most relevant temperature indicator could depend, among other things, on physical activity and physiology, and that age would be a structuring parameter in this case.

240 The most significant temperature indicator, as determined by Akaike's criteria, was then used in all the 241 analyses described below.

242 2.5.3. Mortality risks by causes, age, sex and territorial sub-division

We investigated the temperature-mortality relationship by causes, age, sex, and type of territory, based on the previously selected temperature indicator. We used stratified models, based on the assumption that age and sex were key parameters for the constituent of vulnerability to heat (the individual susceptibility to heat, the ability to cope with heat, and the ability to adapt). We additionally stratified by type of territory, considering that the exposure patterns and the social mechanisms implied in vulnerability could vary between rural and urban settings.

249 2.5.4. Effect-modification by vegetation and socio-economic indicators

Here, we investigated how socio-economic or vegetation indicators modified the temperature-mortality relationship. We build on the models previously developed, and added a linear interaction between the effect modifier (vegetation or socio-economic indicator) and the temperature term. The results compared the temperature-mortality relationship in municipalities with a low level of the studied indicator (5th percentile) versus municipalities with a high level (95th percentile).

Again, these models were stratified by age, sex, and territorial sub-division.

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258 3. Results

The study area comprised 1,287 municipalities with a total population of approximately 9.8 million individuals aged 15 and over. From 2000 to 2017, there were 140,047 deaths among individuals aged 15-64 and 465,180 deaths among those aged 65 years and over (Table 1). The number of deaths due to cardiovascular, cardiac, and respiratory mortality was insufficient to develop models by territorial sub-division for the age group 15-64.

Table 1 – Population (2017) and mortality data (2000-2017) by causes, sex, age-group, and territorial subdivision

			Γ	len		Women							
		Rural areas, small and me- dium- sized cities	Parisian agglo- meration	Hypercen- ter and inner met- ropolitan area	Total	Rural areas, small and me- dium- sized cities	Parisian agglo- meration	Hyper- center and inner metro- politan area	Total				
Population	15-64	486,903	1,075,217	2,341,468	3,903,588	498,165	1,115,865	2,492,749	4,106,779				
	65+	109,326	204,512	446,199	760,037	138,403	266,499	620,230	1,025,132				
All-cause mortality	15-64	13,188	23,678	54,078	90,944	6,354	12,183	28,306	46,843				
	65+	30,972	52,366	127,245	210,583	35934	60,884	158,847	255,665				
Cardiovas- cular morta- lity	15-64	1,875*	3,495	7,107	12,477	604	1,158	2,439	4,201				
Cardiac mor-	00+	8,060	13,270	31,400	52,604	10,469	17,227	43,822	71,536				
tality	15-64	1,387	2,547	5,101	9,035	372	697	1,507	2,576				
	65+	5,804	9,262	22,260	37,326	7,339	12,035	30,408	49,782				
Respiratory mortality	15-64	302	569	1,241	2,112	123	243	572	938				
-	65+	2,310	3,735	8,866	14,911	2,342	3,774	9,836	15,952				

266 *values in italics indicate numbers that are too small for modelling purposes (<5,000)

The mean daily temperature, averaged over extended summers (April-September) for the period 268 2000-2017 varied between 16.1°C in rural areas and small to medium-sized cities, and 17.4°C in the 269 hyper-centre and inner metropolitan area. Warmer minimum temperatures were observed in the densest urban areas, while maximum temperature were more heterogeneous through the region (seeFigure S1 for an illustration).

The percentage of artificialized area was significantly lower in rural, small, and medium-sized cities (9.9%) compared to other areas. However, the distribution of NDVI and the percentage of surface lacking tree canopy were similar between rural areas, small and medium-sized cities, and the Parisian agglomeration (e.g. 0.57 and 0.55 for NDVI, respectively). Figure 3 and Figure S2 indicate a lack of vegetation in the hyper-centre and inner metropolitan area compared to other areas.

The percentage of social housing varied across different areas, with rural areas and small to mediumsized cities having a lower percentage (2.9%) compared to the hyper-centre and inner metropolitan area (25%). The median income in the hyper-centre and inner metropolitan area was generally lower, with larger differences between municipalities compared to other areas. Social deprivation was also more heterogeneous in the hyper-centre and inner metropolitan area, as shown in Figure 3 and Figure S3.

Mean temperature was strongly correlated with all other temperature indicators (Supplementary Figure S5). Supplementary Figure S6 details the correlation between mean summer temperature indicators, and vegetation and socio-economic indicators. Mean temperature showed a correlation with % of social housing (0.47), NDVI (-0.47), and artificialized areas (0.62) over the entire region. However, it did not show any correlation with lack of tree canopy (0.08). In urban areas, minimum temperature was correlated with NDVI and artificialized area.

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299 Figure 3 – Distribution of the environmental and socio-economic indicators by territorial sub-





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302 3.2. Selection of the temperature indicator

303 These analyses were limited to the entire region. For all temperature indicators, a similar J-shaped 304 mortality response was found, with a significant increase in relative risks (RR) at extreme

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temperatures (supplementary material Figure S7). Table 2 displays RR for each indicator for all-cause
 mortality, for high temperature percentiles. Supplementary material Table S1 provides details for
 cardiovascular, cardiac, and respiratory mortality.

308 Supplementary Figure S8 provides information on the response by lag for selected temperature 309 percentiles. The impact of minimum daily temperature was limited to the first 48 hours after exposure 310 and lasted a maximum of 6 days at percentile 99.9, while the effect of maximum daily temperature 311 persisted for 4 to 10 days.

Mean daily temperature was identified as the best indicator for modelling all causes of mortality, for both age groups, based on the Akaike criteria. However, the Akaike criteria showed that all indicators were very similar (see supplementary material Table S2). Therefore, mean daily temperature was used for the analyses presented below.

316 Table 2 – Relative risks (RR) (0-10 lags) and 95% confidence intervals for all-cause mortality by

317 age group, for selected percentiles of each daily temperature indicators calculated over the

318 whole region

	Age	P95 vs P50	P99 vs P50	P99.5 vs P50	P99.9 vs P50
Mean daily		23.2°C vs 16.6°C	26.3 °C vs 16.6°C	27.3°C vs 16.6°C	29.4°C vs
tomporaturo					16.6°C
temperature	15-64	0.97 [0.93:1.00]	1.12 [1.06:1.17]*	1.27 [1.21:1.34]	1.95 [1.79:2.13]
	65+	0.93 [0.91:0.94]	1.21 [1.18:1.24]	1.53 [1.49:1.57]	3.23 [3.10:3.37]
Maximal daily		30.7°C vs 21.8°C	34.6°C vs 21.8°C	36.2°C vs 21.8°C	39.1 °C vs
temperature					21.8°C
-	15-64	0.99 [0.95:1.03]	1.22 [1.15:1.30]	1.52 [1.43:1.63]	2.77 [2.39:3.21]
	65+	0.96 [0.94:0.99]	1.45 [1.40:1.49]	2.20 [2.13:2.28]	6.95 [6.48:7.45]
Minimal daily		17.3°C vs 11.3°C	19.3°C vs 11.3°C	20.1°C vs 11.3°C	21.7 °C vs
temperature					11.3°C
·····	15-64	0.94 [0.91:0.97]	1.02 [0.97:1.06]	1.11 [1.06:1.17]	1.54 [1.44:1.65]
	65+	0.89 [0.87:0.90]	1.01 [0.99:1.04]	1.18 [1.15:1.20]	1.95 [1.89:2.02]
Mean daily radiant		34.8°C vs 25.9°C	37.9°C vs 25.9°C	39.0°C vs 25.9°C	40.5°C vs 25.9°C
temperature	15-64	0.99 [0.94:1.04]	1.29 [1.21:1.37]	1.58 [1.45:1.73]	2.34 [1.97:2.78]
	65+	1.13 [1.10:1.16]	1.81 [1.75:1.87]	2.50 [2.39:2.61]	4.48 [4.12:4.87]

Mean daily UTCI		26.2°C vs 17.6°C	29.7°C vs 17.6°C	31.1°C vs 17.6°C	33.3°C vs 17.6°C
	15-64	0.97 [0.93:1.01]	1.21 [1.14:1.27]	1.49 [1.40:1.59]	2.54 [2.24:2.89]
	65+	0.96 [0.94:0.98]	1.47 [1.43:1.51]	2.17 [2.10:2.24]	5.64 [5.30:6.00]

319 * significant RR are presented in bold

320 3.3. Mortality risks by causes, age, sex and territorial sub-divisions

These models examined the association between heat and all-cause mortality by age, sex and type of territories using the mean daily temperature. The results showed that heat was consistently associated with an increased risk of all-cause mortality, and that small changes in temperature resulted in large changes in RR (Table 3). The increase in RR was steeper for people aged 65 years and over.

Significant RR for women aged 15-64 were only observed in the hyper-centre and inner metropolitan area, while significant RR for men were also observed in rural, small and medium-size cities, and in the Parisian agglomeration. For men aged 15-64 living in the hyper-centre and inner metropolitan area, at a mean temperature of 29.4°C (99.9th percentile) the RR was 2.17 [1.93:2:45]. The RR for women of the same age group in the same area was 1.76 [1.48:2.08].

330 By contrast, for people aged 65 and over, larger RR were observed for women in all areas, and the 331 difference with men increased with temperature. In this age group, the larger RR for women were 332 observed in the rural, small and medium-size cities. At a mean temperature of 29.4°C (99.9th 333 percentile), the relative risk (RR) for men aged 65 years and over living in rural areas, small and 334 medium-sized cities was 1.89 [1.39:2.57]. The RR for women of the same age group in the same area 335 was 3.94 [3.09:5.01]. The RR for men aged 65 and over living in the hyper-centre and inner 336 metropolitan area was 2.44 [2.26:2:64], while the RR for women of the same age group in the same 337 area was 3.44 [3.24:3.66]. Similar patterns were observed for cardiovascular, cardiac and respiratory 338 mortality in people aged 65 years old and over. The largest RR were observed for respiratory 339 mortality: in the hyper-centre and inner metropolitan area, a temperature of 26.3°C was associated 340 with a RR of 1.39 [1.14:1.70] in men aged 65 and over, and a RR of 1.55 [1.29:1.87] in women aged 65 and over. At a temperature of 29.4°C, the RR were respectively 3.01 [2.29:3.96] and 4.12 341 342 [3.23:5.25].

343 RR stratified by territorial sub-division and age are presented in Supplementary material Table S3 and

344 S4.

345

346Table 3 – Relative risk (RR) (0-10 lags) and 95% confidence interval for selected percentiles of347mean temperature, all-cause mortality, by age and sex and territorial sub-division - RR in

348 reference to P50 (16.6°C)

		All-cause mortality										
			Men			Women						
	Age	P99 26.3 °C	P99.5 27.3°C	P99.5 27.3°C	P99 26.3 °C	P99.5 27.3°C	P99.5 27.3°C					
Rural areas, small and medium- sized cities Parisian	15-64	1.17 [0.98:1.39]	1.27 [1.02 :1.57]	1.59 [1.01:2.51]	1.06 [0.82:1.38]	1.17 [0.85:1.61]	1.57 [0.78:3.13]					
	65 +	1.13 [1.00:1.27]	1.28 [1.11:1.47]	1.89 [1.39:2.57]	1.47 [1.33:1.63]	1.87 [1.67:2.11]	3.94 [3.09:5.01]					
Parisian agglomeration	15-64	1.03 [0.91:1.17]	1.14 [1.00:1.31]	1.64 [1.26:2.15]	0.98 [0.83:1.17]	1.07 [0.89:1.30]	1.45 [0.99:2.14]					
	65 +	1.18 [1.09:1.29]	1.40 [1.28:1.54]	2.41 [2.03:2.87]	1.29 [1.19:1.40]	1.68 [1.54:1.82]	3.90 [3.37:4.51]					
Hyper-centre and inner metropolitan	15-64	1.12 [1.04:1.21]	1.29 [1.19:1.40]	2.17 [1.93:2.45]	1.05 [0.94:1.18]	1.17 [1.05:1.31]	1.76 [1.48:2.08]					
area	65 +	1.07 [1.01:1.12]	1.27 [1.20:1.34]	2.44 [2.26:2.64]	1.13 [1.08:1.19]	1.43 [1.37:1.50]	3.44 [3.24:3.66]					
			Mortalit	y by causes, (65 years old a	and over						
			Mortalit Men	y by causes, (l 65 years old a	and over Women						
	Causes	P99 26.3 °C	Mortalit Men P99.5 27.3°C	y by causes, 6 P99.5 27.3°C	65 years old a P99 26.3 °C	Women P99.5 27.3°C	P99.5 27.3°C					
Rural areas, small and medium-	Causes Cardiovascular	P99 26.3 °C 1.04 [0.83:1.31]	Mortalii Men P99.5 27.3°C 1.22 [0.93:1.61]	P99.5 27.3°C 2.01 [1.11:3.63]	P99 26.3 °C 1.45 [1.21:1.76]	P99.5 27.3°C 1.79 [1.44 :2.23]	P99.5 27.3°C 3.35 [2.13:5.27]					
Rural areas, small and medium- sized cities	Causes Cardiovascular Cardiac	P99 26.3 °C 1.04 [0.83:1.31] 1.06 [0.81:1.39]	Mortalii Men P99.5 27.3°C 1.22 [0.93:1.61] 1.16 [0.83:1.63]	P99.5 27.3°C 2.01 [1.11:3.63] 1.51 [0.73:3.15]	P99 26.3 °C 1.45 [1.21:1.76] 1.56 [1.24:1.95]	P99.5 27.3°C 1.79 [1.44 :2.23] 1.98 [1.52 :2.58]	P99.5 27.3°C 3.35 [2.13:5.27] 4.06 [2.35:6.99]					
Rural areas, small and medium- sized cities Parisian agglomeration	Causes Cardiovascular Cardiac Cardiovascular	P99 26.3 °C 1.04 [0.83:1.31] 1.06 [0.81:1.39] 1.18 [1.00:1.40]	Mortalia Men P99.5 27.3°C 1.22 [0.93:1.61] 1.16 [0.83:1.63] 1.35 [1.13:1.63]	P99.5 27.3°C 2.01 [1.11:3.63] 1.51 [0.73:3.15] 2.04 [1.44:2.92]	P99 26.3 °C 1.45 [1.21:1.76] 1.56 [1.24:1.95] 1.25 [1.08:1.44]	Nomen P99.5 27.3°C 1.79 [1.44 :2.23] 1.98 [1.52 :2.58] 1.51 [1.29 :1.77]	P99.5 27.3°C 3.35 [2.13:5.27] 4.06 [2.35:6.99] 2.78 [2.11 :3.67]					
Rural areas, small and medium- sized cities Parisian agglomeration	Causes Cardiovascular Cardiac Cardiovascular Cardiovascular	P99 26.3 °C 1.04 [0.83:1.31] 1.06 [0.81:1.39] 1.18 [1.00:1.40] 1.18 [0.97:1.44]	Mortalia Men P99.5 27.3°C 1.22 [0.93:1.61] 1.16 [0.83:1.63] [1.13:1.63] [1.13:1.63] 1.37 [1.10:1.70]	P99.5 27.3°C 2.01 [1.11:3.63] 1.51 [0.73:3.15] 2.04 [1.44:2.92] 2.16 [1.42:3.27]	P99 26.3 °C 1.45 [1.21:1.76] 1.56 [1.24:1.95] 1.25 [1.08:1.44] 1.28 [1.07:1.53]	Momen P99.5 27.3°C 1.79 [1.44 :2.23] [1.52 :2.58] [1.52 :1.77] 1.50 [1.24 :1.81]	P99.5 27.3°C 3.35 [2.13:5.27] 4.06 [2.35:6.99] 2.78 [2.11:3.67] 2.45 [1.74:3.46]					
Rural areas, small and medium- sized cities Parisian agglomeration Hypercenter and inner metropolitar	Causes Cardiovascular Cardiovascular Cardiovascular Cardiovascular Cardiovascular	P99 26.3 °C 1.04 [0.83:1.31] 1.06 [0.81:1.39] 1.18 [1.00:1.40] 1.18 [0.97:1.44] 0.98 [0.88:1.10]	Mortalia Men P99.5 27.3°C 1.22 [0.93:1.61] 1.16 [0.83:1.63] 1.35 [1.13:1.63] 1.37 [1.10:1.70] 1.13 [1.01:1.26]	P99.5 27.3°C 2.01 [1.11:3.63] 1.51 [0.73:3.15] 2.04 [1.44:2.92] 2.16 [1.42:3.27] 1.90 [1.61:2.23]	P99 26.3 °C 1.45 [1.21:1.76] 1.25 [1.08:1.44] 1.28 [1.07:1.53] 1.13 [1.03:1.23]	Women P99.5 27.3°C 1.79 [1.44 :2.23] [1.52 :2.58] [1.52 :2.58] [1.29 :1.77] [1.24 :1.81] 1.36 [1.24 :1.49]	P99.5 27.3°C 3.35 [2.13:5.27] 4.06 [2.35:6.99] 2.78 [2.11 :3.67] 2.45 [1.74 :3.46] 2.68 [2.36:3.04]					

	[0.86:1.11]	[1.00:1.30]	[1.72:2.51]	[1.06 :1.31]	[1.30 :1.62]	[2.60:3.51]
Respiratory	1.39 [1.14:1.70]	1.66 [1.35:2.03]	3.01 [2.29:3.96]	1.55 [1.29:1.87]	1.94 [1.60:2.35]	4.12 [3.23:5.25]

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351 **3.4. Effect modification by vegetation and socio-economic indicators, by age and sex, for the** 352 whole region

353 a) Vegetation indicators

354 In the entire region, people aged 15-64 residing in municipalities with more vegetation (as determined by the percentage of artificialized areas, NDVI, or lack of tree canopy) had a lower risk of heat-related 355 356 mortality compared to those living in municipalities with less vegetation (Figure 4). The difference in 357 risk increased with temperature. However, this protective effect of vegetation was not observed among 358 individuals aged 65 years and above. For example, when the temperature was 29.4°C compared to 16.6°C, the RR for all-cause mortality in the 15-64 age group was 2.04 [1.83:2.29] in a municipality 359 360 with 98% of its surface area lacking tree canopy, compared to 1.31 [1.04:1.82] in a municipality with 361 51% of its surface area lacking tree canopy. In the 65 and over age group, the RR were 3.11 [2.95:3.27] and 3.13 [2.78:3.52], respectively. 362

363 b) Socio-economic indicators

For both age groups, the risk of mortality due to heat was higher in municipalities with higher levels of 364 365 deprivation (higher FDep), a higher percentage of social housing, or a lower median income (Figure 366 4). For example, among people aged 15-64 years old, the RR associated with a mean temperature of 367 29.4°C compared to 16.6°C was 2.08 [1.88:2.30] in a municipality with 34% social housing, compared 368 to 1.63 [1.38:1.93] in a municipality without social housing. For individuals aged 65 and over, the RR 369 were 3.38 [3.21:3.56] and 2.96 [2.76:3.19], respectively. However, when stratifying by sex, this effect 370 of socio-economic indicators was observed in women only. Women aged 65 and over residing in the 371 most deprived municipalities according to the FDep had an RR at 29.4°C of 4.16 [3.83:4.51], 372 compared to 3.44 [3.20:3.71] for women living in less deprived municipalities. For men, the RR were 373 respectively 2.55 [2.30:2.83] and 2.43 [2.20:2.68].

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- 382 Figure 4 RR for all-cause mortality by vegetation and socio economic factors, for a mean

383 temperature at P99.9 vs P50 (29.4°C, vs 16.6°C) by age-group and sex – entire region

	All-cause mortality	/ 15-64	
Effect	Metric		RR (95% CI)
Artificialized area			
	More artificialized (P95 93.2%)		2.02 [1.84:2.21]
	Less artificialized (0%)	-	1.37 [1.01:1.87]
NDVI			
	Less vegetation (P5 0.35)		1.85 [1.68:2.03]
	More vegetation (P95 0.71)	-	1.44 [1.07:1.94]
Lack of tree canopy			
	Less trees (P95 98%)		2.04 [1.83:2.29]
	More trees (P5 51%)	_ 	1.31 [1.04:1.82]
Social housing			
	More social housing (P95 34%)		2.08 [1.88:2.3]
	Less social housing (P5 0%)		1.63 [1.38:1.93]
Income			
	Lower income (P5 17,858€)		2.03 [1.83:2.25]
	Higher income (P95 35,106€)		1.75 [1.52:2.01]
FDep			
	More deprived (P95 0.77)		- 2.16 [1.9:2.44]
	Less deprived (P5 -3.16)		1.72 [1.49:1.99]
		1.5 2	

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			Whole population 65+						Men 65+					Women 65+			
Effect	Metric	Percentile					RR (95% CI)					RR (95% CI)					RR (95% CI)
Artificialized area								1									
	More artificialized	P95 (93%)		٠			3.23 [3.09:3.38]		-			2.55 [2.37:2.74]			+		3.69 [3.49:3.91]
	Less artificialized	P5 (0%)					3.02 [2.64:3.46]	1				1.96 [1.56:2.46]				_	3.93 [3.32:4.65]
NDVI																	
	Less vegetation	P5 (0.35)	-	+			3.25 [3.11:3.4]		+			2.44 [2.27:2.62]			-		3.86 [3.64:4.09]
	More vegetation	P95 (0.71)	-		-		3.42 [2.98:3.91]	1				2.11 [1.69 :2.63]			_	•	4.69 [3.95 :5.57]
Tree canopy																	
	Less trees	P95 (98%)		٠			3.11 [2.95:3.27]		-			2.52 [2.31:2.74]			-		3.51 [3.28:3.76]
	More trees	P5 (51%)					3.13 [2.78:3.52]					2.06 [1.69:2.51]				_	4.00 [3.45:4.63]
% social housing								1									
	More social housing	P95 (34%)	-	+			3.38 [3.21:3.56]	1	+			2.53 [2.33 :2.75]			-+-		4.01 [3.75 :4.28]
	Less social housing	P5 (0%)	-	+			2.96 [2.76:3.19]	1	-			2.41 [2.14:2.71]		-	•		3.31 [3.02:3.63]
Income																	
	Lower income	P5 (17,858 €)		+			3.31 [3.14:3.49]		+			2.50 [2.29:2.72]			-		3.89 [3.64:4.16]
	Higher income	P95 (35,106 €)		•			3.12 [2.95:3.29]					2.46 [2.25:2.70]					3.56 [3.32:3.82]
FDep								1									
	More deprived	P95 (0.77)		+			3.46 [3.25:3.69]					2.55 [2.30:2.83]			-+	-	4.16 [3.83:4.51]
	Less deprived	P5 (-3.16)		+			3.04 [2.86:3.22]		-			2.43 [2.20:2.68]			•		3.44 [3.20:3.71]
			1 2	3	4	5	6	1	2 3	4	5	6	1 :	2 3	4	5	6

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386 3.5. Effect modification by vegetation and socio-economic indicators, by age and sex, and

387 territorial sub-division

388 a) Vegetation indicators

389 When stratifying the analysis by territorial sub-division, the protective effect of vegetation indicators 390 disappeared for people aged 15-64 years old in all areas. For people aged 65 and over, vegetation 391 indicators did not modify the RR in rural areas, small and medium-sized cities. In the Parisian 392 agglomeration and in the hypercenter and inner metropolitan area, RR were higher in municipalities 393 with higher vegetation. For instance, in the Parisian agglomeration, the RR at 29.4°C was 2.85 394 [2.46:3.30] in municipalities with 89% of artificialized surfaces compared to 4.59 [3.56:5.92] in 395 municipalities with 49% of artificialized surfaces (Figure 4). Differences by sex were again observed: in 396 rural areas, small and medium size-cities, men living in municipalities with more vegetation tended to 397 have higher RR, but the opposite was observed for women. In the Parisian agglomeration, and in the 398 hypercenter and inner metropolitan area, women living in municipalities with more vegetation tended 399 to have higher RR (Supplementary material Table S5). Results by sex are however difficult to interpret 400 due to the large confidence intervals.

401 b) Socio-economic indicators

402 For both age groups, in the hypercenter, RR were higher in municipalities with more social housing. 403 For instance, for people 65 years old and over, in the hyper-center and inner metropolitan area, a 404 temperature at P99.9 was associated with an all-cause mortality RR of 3.35 [3.03:3.70] in 405 municipalities with 18% of social housing, compared to 2.87 [2.68:3.08] in municipalities with 6% social 406 housing (Figure 4). In rural areas, small and medium-sized cities, socio-economic indicators did not 407 modify the results by sex in people aged 65 years old and over. In the Parisian agglomeration, men 408 living in municipalities with more social housing, or with higher social deprivation, tended to have 409 larger RR, but the opposite was observed for women. In the hypercenter and inner metropolitan area, 410 no difference was observed for men, while women in municipalities with more social housing, or with 411 higher social deprivation, tended to have larger RR (Supplementary Table 5).

412 **3.6. Sensitivity analyses**

Results were not modified when including a spatial component in the model to account for spatialautocorrelation and heterogeneity (data not shown).

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422 Figure 4 - Relative risk (RR) (0-10 lags) by age, all-cause mortality for a mean temperature at

423 **P99.9 vs P50 (29.4°C, vs 16.6°C)**

		Rural areas, small and medium-sized cities						Parisia	in agglomerati	on	Hypercenter and inner metropolitan area				
Effect	Metric	Percentile				RR (95% CI)	Percentile			RR (95% CI)	Percentile			RR (95% CI)	
Artificialized area			1					1							
15-64															
	More artificialized	P95 (35%)				1.55 [1.06:2.25]	P95 (89%)			1.44 [1.07:1.95]	P95 (100%)			2.04 [1.82:2.28]	
	Less artificialized	P5 (0%) -	•	-		1.45 [0.81:2.60]	P5 (13%)			1.84 [1.08:3.13]	P5 (49%)			1.90 [1.47:2.46]	
65+															
	More artificialized	P95 (35%)	1	-	-	2.96 [2.45:3.59]	P95 (89%)			2.85 [2.46:3.30]	P95 (100%)	-		2.99 [2.83:3.16]	
	Less artificialized	P5 (0%)	-	•	-	2.76 [2.08:3.67]	P5 (13%)			- 4.59 [3.56:5.92]	P5 (49%)	-•		3.16 [2.85:3.20]	
Lack of tree canopy															
15-64															
	Less trees	P95 (99%)				1.70 [0.93:3.12]	P95 (93%)			1.56 [1.11:2.19]	P95 (97%)			2.02 [1.80:2.27]	
	More trees	P5 (51%) —	•	-		1.27 [0.62:2.59]	P5 (46%)	+	-	1.44 [0.83:2.49]	P5 (63%)			1.87 [1.41:2.49]	
65+															
	Less trees	P95 (99%)		•		2.89 [2.15:3.90]	P95 (93%)			2.97 [2.51:3.51]	P95 (97%)			2.94 [2.78:3.12]	
NDV	More trees	P5 (51%)		•	_	2.91 [2.10:4.02]	P5 (46%)			3.57 [2.82:4.52]	P5 (63%)		•	3.40 [3.01:3.83]	
15-64	1	DE (0.44)				1 40 10 00-0 061	DE (0.42)			4 62 (4 00-4 00)	DE (0.46)	_		1 00 14 64-0 101	
	Less vegetation	P5 (0.41)				1.49 [0.96:2.26]	P5 (0.42)	-		1.55 [1.22:1.92]	P5 (0.16)			1.90 [1.64:2.19]	
65+	wore vegetation	P95 (0.72) -				- 1.66 [0.70:5.02]	F95 (0.70)	_		1.02 [0.40:2.27]	P95 (0.55)			2.20[1.74.2.92]	
0.0+	Loss vogotation	P5 (0.41)			_	3 04 [2 43:3 72]	P5 (0 42)			3 14 [2 70:3 55]	P5 (0.16)	-		2 80 [2 61-3 01]	
	More vegetation	P95 (0.72)				2 61 [1 63:4 18]	P95 (0.70)				P95 (0.53)	_		- 3 77 [3 33:4 26]	
Income	wore vegetation	1 33 (0.72)				2.01[1.00.4.10]	1 33 (0.10)		-	4.00 [0.00.0.00]	1 33 (0.35)			0.77 [0.00.4.20]	
15-64															
	Lower income	P5 (19.882€)				1.54 [1.04:2.27]	P5 (16.092€)			1.71 [1.27:2.30]	P5 (12.080€)			2.22 [1.91:2.59]	
	Higher income	P95 (34,248€) —				1.29 [0.56:2.98]	P95 (37,136€)			1.23 [0.73:2.08]	P95 (38.655€)			1.75 [1.48:2.07]	
65+							(
	Lower income	P5 (19,882€)			-	2.94 [2.41:3.58]	P5 (16,092€)			2.88 [2.45:3.40]	P5 (12,080€)	-	-	3.26 [3.01:3.53]	
	Higher income	P95 (34,248€)		•		2.97 [1.98:4.45]	P95 (37,136€)			3.94 [3.16:4.92]	P95 (38,655€)			2.84 [2.64:3.05]	
% Social housing															
15-64															
	More social housing	P95 (18%)				1.52 [1.05:2.22]	P95 (38%)		-	1.75 [1.28:2.38]	P95 (51%)			2.31 [1.92:2.80]	
	Less social housing	P5 (0%) -				1.38 [0.79:2.41]	P5 (0%)	֥		1.25 [0.80:1.94]	P5 (6%)			1.81 [1.55:2.11]	
65+															
	More social housing	P95 (18%)			-	2.96 [2.45:3.58]	P95 (38%)			3.28 [2.74:3.93]	P95 (51%)		•	3.35 [3.03:3.70]	
	Less social housing	P5 (0%)		•	_	2.88 [2.20:3.78]	P5 (0%)			3.22 [2.60:3.98]	P5 (6%)			2.87 [2.68:3.08]	
FDep															
15 - 64			1												
	More deprived	P95 (0.63		-		1.67 [1.07:2.61]	P95 (0.72)		-	1.74 [1.27:2.38]	P95 (1.61)			2.40 [2.01:2.85]	
	Less deprived	P5 (-2.84 —	-	_		1.10 [0.43:2.78]	P5 (-3.44)			1.20 [0.69:2.09]	P5 (-3.92)			1.67 [1.39:2.00]	
65+															
	More deprived	P95 (0.63	1		-	3.04 [2.43:3.81]	P95 (0.72)			2.99 [2.52:3.54]	P95 (1.61)	-	•	3.41 [3.11:3.74]	
	Less deprived	P5 (-2.84		-	_	2.76 [1.78:4.27]	P5 (-3.44)			3.77 [2.95:4.80]	P5 (-3.92)			2.75 [2.54:2.98]	
			1 2	3	4	5		1 2	3 4 5	6		1 2 3	4		

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425 4. Discussion

426 **4.1. Summary of the main findings**

427 This study confirms that heat is associated with an increased risk of mortality in the Paris region. It is

428 the first study in the Paris region to differentiate risks by age, sex, and territorial sub-division. Mortality

429 risks were of a similar order of magnitude between all territories. However, for the highest temperature

430 percentiles, the risks became larger in the hypercenter and inner metropolitan area.

The study suggested that women aged 65 and over are at higher risk than any other population of the region. The results also found that vegetation had a protective effect in the 15-64 age group only. For those aged 65 and over, sex and socio-economic factors appeared to be a major determinant of heatrelated mortality.

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436 4.2. Heat-related risks by territorial sub-division

437 Heat is usually considered as a risk in urban areas. There have been limited studies on its impact in 438 rural areas, and almost none in developed countries or Europe (Odame, Li et al. 2018). The literature 439 suggests that heat-related risks in rural areas are similar in magnitude to those in urban settings 440 (Odame, Li et al. 2018). In the Paris Region, similar heat-related risks were observed in rural areas, 441 small and medium-sized cities, and in the densest urban areas of the region, for all causes and all age 442 groups. Differences only appeared at very high temperature percentiles, when the RR were higher in 443 the hypercenter and inner metropolitan area confirming the importance of short-term adaptation of 444 these territories. Our results indicate the need for a stronger adaptation to heat in rural areas, small 445 and medium-sized cities in the Paris region. In addition, in the coming years, rural areas may warm 446 faster than urban areas, potentially resulting in higher mortality burden in those areas (Lemonsu, 447 Kounkou-Arnaud et al. 2013, Le Roy, Lemonsu et al. 2020, Michau Y, Lemonsu A et al. 2024).

To go further in analysing the differences between urban, suburban and rural territories, it would be relevant to reproduce similar analyses in other regions, and combine the results in a meta-regression.

450 **4.3. Heat-related risks by causes of deaths**

Our study showed that heat increased all-cause, cardiovascular, cardiac and respiratory mortality, consistently with current knowledge (Cheng, Xu et al. 2019). The literature reports significant adverse impacts of heat on cardiovascular (Gasparrini, Armstrong et al. 2012, Cheng, Xu et al. 2019, Alahmad, Khraishah et al. 2023), ischemic (Gasparrini, Armstrong et al. 2012, Alahmad, Khraishah et al. 2023, Zhang, Breitner et al. 2023), cerebrovascular (Gasparrini, Armstrong et al. 2012, Conti, Valente et al. 2022, Alahmad, Khraishah et al. 2023), and respiratory (Gasparrini, Armstrong et al. 2012, Cheng, Xu et al. 2019, Zafeiratou, Samoli et al. 2023) mortality mortality during heat waves, or during warm days.

458 Other causes of mortality such as renal mortality, diabetes, injuries, and mental health related-459 mortality including suicides can also be affected by heat and heat waves (Gasparrini, Armstrong et al. 460 2012) (Conti, Valente et al. 2022), and should be included in future analyses if sufficient number of 461 cases were available. Other health impacts such as emergency room visits, hospital admissions, or 462 sleep disturbance should also be explored to have a better understanding of the overall heat-health 463 impacts.

464 Finally, this study focused on the short-term impacts of temperature only, but long-term impacts could465 also result in impaired health (Zafeiratou, Samoli et al. 2021).

466 **4.3. Heat-related risks by age and sex**

At moderate temperatures, the risks for younger men were larger than those for older men. This may indicate a risk linked with occupational exposure or physical activity in men. It is also possible that risk perceptions and adoption of protective behaviours are lower among young men.

470 In the older age group, by contrast, RR were larger for women compared to men. The literature reports 471 differences in heat-related mortality risk between men and women, but with few insights on the 472 possible drivers. A literature review focusing on 13 European studies published between 2000 and 473 2016 reported that the majority (8 studies) found higher mortality risks in elderly women (van Steen, 474 Ntarladima et al. 2019). Similar results were reported in more recent studies, for instance with higher 475 risks in women reported in Belgium (all ages) (Demoury, Aerts et al. 2022). Slightly higher RR were 476 reported for men younger than 65 years old in the Czech republic, similar RR between men and 477 women between 64 and 84 years old, and higher RR for women older than 84 years old (Vésier and 478 Urban 2023). In the Netherlands, the RR were higher for men in the youngest age group (<65), and 479 higher for women in older age groups (Folkerts, Bröde et al. 2022).

Sex and gender can influence physiological, behavioural and social risk factors (Folkerts, Bröde et al. 2022). A review of sex differences in thermoregulation for young fit women involved in physical activities concluded that, despite physiological differences, there was no evidence that women have an inherent susceptibility to heat-related illnesses (Yanovich, Ketko et al. 2020). Elderly women may be more susceptible to heat due to hormonal changes after the menopause, but the knowledge on this remains scarce (van Steen, Ntarladima et al. 2019).

Age can also partly explain the differences, as women are over-represented in the eldest age groups.
In 2020, in the Paris region, among women aged 65 and over, 51% were aged 65-74 years old, 41%

488 75-89 years old, and 8% were 90 and over. Among men aged 65 and over, 58% were aged 65-74 489 years old, 38% 75-89 years old, and 4% were 90 and over (Insee 2020). However, based on their 490 literature review, van Steen *et. al.* concluded that age alone cannot explain all the differences in heat-491 related risks observed between men and women (van Steen, Ntarladima et al. 2019).

492 In France, gender is a significant factor in driving inequities. In 2018, women earned on average 20-493 25% less than men. Among retired individuals, women's income was 29% lower than men's, and up to 494 53% lower in the poorest socioeconomic category (Insee 2021). In the city of Paris, women aged 50 495 years and over earn on average 32% less than men in the same age group. Additionally, the poverty 496 rate has increased more rapidly among women, particularly those aged 75 years and older (Insee 497 2021). Women aged 75 and over are more likely to experience isolation than men (Observatoire des 498 inégalités 2019). Women aged 85 years and over are also more likely to live in nursing homes (11% 499 compared to 7% of men), to experience social isolation at home (55% compared to 24% of men), and 500 to live with relatives less often than men (34% compared to 69% of men) (Insee 2022). Reducing sex 501 inequities and social isolation could be key to reduce the heat impacts in women.

502 **4.4. Influence of vegetation and socio-economic indicators on the temperature-mortality** 503 **relationship**

504 When considering the entire region, municipalities with less artificialized surfaces, higher NDVI, higher 505 tree covering, lower number of social housing or higher incomes had lower all-cause and 506 cardiovascular mortality heat-related risks. When stratifying the results by territorial sub-division, we 507 found that artificialized areas and NDVI were no longer associated with a protective effect.

508 Several studies worldwide reported a protective influence of vegetation on the heat-related mortality. A 509 previous study in the Paris region, based from temperature recorded at monitoring stations, found 510 higher heat-related risks in in municipalities with less green spaces, less trees, and more impervious 511 soil (Pascal, Goria et al. 2021). A meta-analysis based on six papers with substantial methodological 512 heterogeneity concluded that the heat-related mortality risk was about 5% higher in less green areas 513 (Schinasi, Benmarhnia et al. 2018). A recent study on heat-related cardiovascular mortality in Norway, Germany and United Kingdom also reported higher risks for women, and for people living in dense 514 515 urban areas with low percentage of green spaces and high levels of particulate matter ($PM_{2.5}$)(Zhang,

516 Breitner et al. 2023). Similar results were obtained for respiratory mortality in the same countries 517 (Zafeiratou, Samoli et al. 2023).

518 As previously mentioned, vegetation can modify temperature, but also exposure and individual 519 susceptibility to heat. Since we are using fine-scale temperature, part of the influence of vegetation on 520 temperature is already captured in the temperature term. Within dense urban areas, vegetation can 521 have an additional cooling effect that is not entirely captured by the temperature data, but with limited 522 geographical spread. This may result in a differentiated risk at a very fine-scale (e.g. neighborhoods, streets), but cannot be observed when using mortality data at the municipality level. Efforts to collect 523 524 high resolution environmental and health data would be beneficial to further understanding the 525 benefits associated with greenspaces in urban areas.

Additionally, in this study we compared the heat-related risks between municipalities with high vs low level of greening. But it may be that those actual differences in vegetation between municipalities are not important enough to modify the heat-related risk (i.e., in dense urban area, the greenest municipalities are not green enough), or that other risk factors are counterbalancing the benefits of green spaces. For instance, a green space not used by the population may help to reduce the ambient temperature locally, but would not contribute to improve the overall health of the population, nor offer shelter to reduce exposure during extreme heat.

533 Greening cities has been proposed as a solution to adapt to climate change. Lungman et al. proposed 534 that in summer (June-August 2015), 71,25% of the heat-related mortality in Paris could have been 535 prevented by increasing the average tree cover density in the city from 5 to 30% (lungman, Cirach et 536 al. 2023). However, a better understanding of the complex pathways between vegetation, heat and 537 health would be needed to design efficient adaptation policies based on vegetation. This in no way 538 detracts from the overall beneficial role of vegetation on health.

Results also underline the importance of socio-economic indicators, with larger RR observed in municipalities with higher share of social housing, or with lower incomes. Socio-economic vulnerability to heat is difficult to capture, that is why we retained three different socio-economic indicators. Those who are deprived may have fewer resources to prepare, respond and adapt to heat. Deprivation is considered to contribute to health inequalities in general due to multiple disadvantages such as housing, education, environmental, working. In particular, housing condition has been reported as a risk factor for higher heat-mortality risk(Gronlund, Berrocal et al. 2015). In 2013, in the Paris region, 9,3% of the population was experiencing housing difficulties, and the social inequities between home owners and tenants, and especially social housing tenants were increasing(Insee 2017).

Finally, our study does not conclude on the best vegetation or socio-economic indicators to consider when developing heat-adaptation policies. Results are being discussed with local stakeholders as part of a wider debate on climate services (Lemonsu, Alessandrini et al. 2024). The discussions show the importance of having a variety of indicators to respond to different contexts and to fuel rich crosssectoral discussions.

553 4.5. Forces and limits

This study assesses heat-related mortality in the Paris region over a 17-year period using fine-scale data from an extensive temperature database. It aims to provide a more comprehensive understanding of the impact of heat on mortality in the region.

557 One strength of the study is the used of fine-scale data to estimate exposure to vegetation and 558 temperature. For vegetation, the 20 m scale allows for a good accuracy when computing the ratio of 559 green surfaces.

The average surface of a municipality in the region is 9 km². Using a resolution of 1 km for the temperature data, and a population-weighted average, allow accounting for sub-municipal variations and giving more weight to inhabited areas. In densely populated areas, this is equivalent to a weighted average based on surface area. In rural areas, this leads to slightly higher temperatures, as the natural areas, which are often cooler but less frequented, weigh less (Forceville, Lemonsu et al. 2024). Previous studies relied on temperature monitored by weather stations, which captured less geographical variation (Pascal, Goria et al. 2021).

This study did not explore directly the influence of other parameters such as wind speed or humidity. Humidity has been identified as an important parameter for human thermoregulation, but, according to previous studies, does not seem to play an important role in the Paris region (Schaeffer, de Crouy-Chanel et al. 2016). This might be due to a climatological context where temperature and humidity are largely correlated, and where high levels of humidity are rarely observed during heat waves (supplementary Figure S9). Another hypothesis is that humidity has little influence on the mechanisms leading to the majority of heat-related causes (Baldwin, Benmarhnia et al. 2023). Wind speed, 574 humidity and radiation were indirectly explored through the use of UTCI as a possible indicator. Based 575 on the Akaïke criteria, UTCI has similar performances than mean, maximum or minimum temperature. 576 UTCI has been proposed to improve the modelling of heat-health related impacts in the built 577 environment(Leroyer, Bélair et al. 2018), but here it did not show added value compared to mean 578 temperature, which is easier to obtain and to communicate.

This study examines the impact of outdoor temperature on mortality. However, it is important to note that individuals spend the majority of their time indoors, and indoor exposure is significantly influenced by age and sex. In France, between 2003 and 2005, men aged 15-59 spent an average of 15 hours per day indoors, while women spent 16 hours per day indoors. Men over 60 spent an average of 17 hours per day indoors, while women over 60 spent 19 hours per day indoors. During the summer, people tend to spend less time indoors (Zeghnoun A and Dor F. 2010).

The study period stopped in 2017 due to the unavailability of mortality data by causes on the more recent years. However, there has been a marked increase in temperatures and particularly in extremes in the Paris region since 2015 (Forceville, Lemonsu et al. 2024). At the same time, there have been a growing number of initiatives to prevent the associated health impacts and adapt to climate change. It is therefore possible that the impact of heat on health and the modifying factors have already changed in the Paris region. Repeating this type of studies at regular intervals would be needed to capture these changes.

Air pollution was not included, due to the unavailability of data over the whole study area. Possible synergies between air pollution (ozone and particulate matter (PM)) and heat were reported by previous studies for all-cause mortality in multicentric studies including the city of Paris (Analitis, De' Donato et al. 2018, Pascal, Wagner et al. 2021) Indication of a joint influence of heat and PM on cardiovascular mortality was also reported (Analitis, De' Donato et al. 2018).

597 **4.6. Implications for public health**

598 The results indicate that heat vulnerability affects individuals of all ages, genders, and living 599 environments. While prevention and interventions often focus on densely populated urban areas and 600 the elderly, it is important to develop specific actions for rural areas and younger populations. 601 Interventions targeting the urban built environment have been effective in reducing heat-related 602 mortality, but should be complemented with measures to address social inequalities and isolation. An 603 experiment conducted in Italy demonstrated that interventions aimed at reducing social isolation 604 among elderly individuals (>80 years old) could decrease excess mortality during heat waves, 605 although not completely eliminate it (Orlando, Mosconi et al. 2021). Additionally, reducing lifelong 606 gender inequalities could also decrease the observed difference in risk between men and women. It is 607 important to note that no single action is sufficient, and coordinated interventions across multiple 608 domains are crucial for any effective adaptation policy.

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