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1 **Title: Mortality Burden of Diurnal Temperature Range and Its Temporal Changes: A**  
2 **Multi-Country Study**

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42 **ABSTRACT**

43 Although diurnal temperature range (DTR) is a key index of climate change, far few studies  
44 have reported the health burden of DTR and its temporal changes at a multi-country scale.  
45 Therefore, we assessed the attributable risk fraction of DTR on mortality and its temporal  
46 variations in a multi-country data set. We collected time-series data covering mortality and  
47 weather variables from 308 cities in 10 countries from 1972 to 2013. The temporal change in  
48 DTR-related mortality was estimated for each city with a time-varying distributed lag model.  
49 Estimates of each city were pooled using a multivariate meta-analysis. The results showed that  
50 the attributable fraction of total mortality to DTR was 2.5% (95% eCI: 2.3-2.7%) over the entire  
51 study period. In overall countries, the attributable fraction has increased from 2.4% (2.1-2.7%)  
52 to 2.7% (2.4-2.9%) between the first and last study years. This study found that DTR has  
53 significantly attributed to mortality in overall countries, and this attributable fraction has  
54 significantly increased overtime in the USA, UK, Spain, and South Korea. Therefore, because  
55 the health burden of DTR is likely to increase in future, countermeasures are needed against  
56 the increase.

57 **Keywords:** Diurnal temperature range, Attributable mortality risk fraction, Time-varying  
58 Effect, Climate Change.

59 **Abbreviations:** Attributable risk fraction (ARF), Distributed Lag Non-linear Model (DLNM).

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## 62 **1. Introduction**

63 Diurnal temperature range (DTR, i.e., the intra-day temperature change) is a **well-known risk**  
64 **factor** of weather-related human health. Numerous studies have described a positive association  
65 between DTR and mortality (Cao et al. 2009; Lim et al. 2015; Tam et al. 2009; Vutcovici et al.  
66 2014; Yang et al. 2013a), **and have reported that people who are elderly, less educated, female**  
67 **or have cardiovascular or respiratory disease are more susceptible to DTR than others (Kan et**  
68 **al. 2007b; Lim et al. 2012a; Yang et al. 2013b). In addition, because the DTR has been reported**  
69 **as an important meteorological indicator closely related with global climate change (Braganza**  
70 **et al. 2004; Kan et al. 2007b; Yang et al. 2013b), an in-depth investigation of the DTR-mortality**  
71 **relationship becomes important as it helps to assess the future health impact of climate change**  
72 **more comprehensively.**

73

74 Biological mechanisms through which a sudden change in absolute temperature might affect  
75 mortality have been described in previous medical and epidemiological studies (Garrett et al.  
76 2009; Garrett et al. 2011; Greenberg et al. 1983; Keatinge et al. 1984a; Martinez-Nicolas et al.  
77 2015; Qiu et al. 2013). Sudden changes in within-day temperature may cause physiological  
78 health problems (Garrett et al. 2009; Garrett et al. 2011); unstable weather or temperature  
79 changes can lead to the onset of cardiovascular events brought on by increased workload. This  
80 can affect the respiratory system by triggering inflammatory nasal responses (Ballester et al.  
81 1997; Carder et al. 2005; Graudenz et al. 2006; Hashimoto et al. 2004; Imai et al. 1999; Luurila  
82 1980). These mechanisms have been suggested as potential causes of increasing human  
83 mortality (Buguet 2007; Guo et al. 2016).

84

85 Based on this biological evidence, previous studies have tried to estimate the risk of DTR on  
86 mortality (Lim et al. 2015; Tam et al. 2009; Vutcovici et al. 2014). However, most previous  
87 studies assessed the risk of DTR using only terms of relative risk (RR), not attributable risk  
88 fraction which can quantify the mortality burden. Furthermore, because a majority of the  
89 previous studies were conducted in single cities or single countries and used statistically  
90 different methods (Kan et al. 2007b; Lim et al. 2012a; Yang et al. 2013b), results of these  
91 studies might have limited applicability to a multi-country scale.

92

93 Most previous studies estimated the risk of DTR on mortality using historical data (Kan et al.  
94 2007b; Lim et al. 2012a) and the estimated impact of DTR was assumed to be consistent  
95 overtime. However, this assumption might not be suitable in predicting the health impacts of  
96 climate change because several factors, including intrinsic biological (e.g., disease/nutrition  
97 status) and extrinsic factors (e.g., forecast and infrastructure improvements, local environment,  
98 or social system conditions), can modify the population's vulnerability to absolute temperature  
99 and rapid temperature change within a day (Gasparrini et al. 2015a; Linares et al. 2014; Wu et  
100 al. 2014). Therefore, it is important to assess temporal change in the DTR-related mortality  
101 relationship to examine whether people are adapted or mal-adapted to DTR.

102

103 In this study, we assess the percent increases in risks and the attributable risk fraction of DTR  
104 for 308 cities of 10 countries. We examine whether the excessive risks and attributable risk  
105 fractions have changed during the study period. We used a Multi-Country Multi-City (MCC)  
106 Collaborative Network to assess the impacts of weather on mortality using a multi-country data  
107 set as referenced in previous papers (Gasparrini et al. 2015a; Gasparrini et al. 2016; Guo et al.  
108 2014; Guo et al. 2016).

109

## 110 **2. Material and methods**

### 111 **2.1. Data**

112 Time-series data covering mortality and weather variables were collected from 385 locations  
113 in 10 countries: Canada (26 cities, 1986-2011), United States (USA) (135 cities, 1985-2006),  
114 Brazil (18 cities, 1997-2011), Colombia (5 cities, 1998-2013), United Kingdom (UK) (10  
115 regions, 1990-2012), Ireland (6 regions, 1984-2007), Spain (51 cities, 1990-2010), Japan (47  
116 prefectures, 1972-2012), South Korea (7 cities, 1992-2010), and Australia (3 cities, 1988-2009).  
117 For convenience of interpretation, the location is described as “city” in this study. The daily  
118 mortality count is the daily count of death for all causes. **If a daily count of all causes of death**  
119 **was not available for a city, then death for non-external causes (ICD-9: 0-799, ICD-10: A00-**  
120 **R99) was used instead.** DTR was chosen as the exposure index, computed from monitoring  
121 stations as the difference between the daily maximum and daily minimum temperatures.  
122 Detailed information regarding data collection is provided in the Supplementary Material (Data  
123 details).

124

### 125 **2.2. First-stage time series model**

126 The first-stage time series model was divided into a two-step procedure. First, a time-series  
127 regression was applied, based on a generalized linear model using a quasi-Poisson distribution  
128 with parameters for: DTR, the day of week, seasonal long-term trend, inter-day temperature  
129 change (the change in mean temperature between two neighboring days), and absolute  
130 temperature. We modeled the DTR-response curve with a linear function and the lag-response  
131 curve with two internal knots placed at equally spaced values on a log scale using natural cubic



132 B-spline with 14 days of lag. Inter-day temperature change was adjusted in the same way as  
133 DTR. We also modeled the temperature-response relationship using a quadratic B-spline with  
134 three internal knots (placed at the 10th, 75th, and 90th percentiles of location-specific  
135 temperature distributions) and a lag-response (up to 21 days) curve with natural cubic B-spline  
136 with three internal knots placed at equally spaced values on the log scale. This model approach  
137 was used in a previous multi-country temperature-mortality study using a distributed lag non-  
138 linear model (DLNM) (Gasparrini et al. 2010; Gasparrini et al. 2015b). Seasonal trends were  
139 adjusted using a natural cubic B-spline of time with 8 degrees of freedom (df) per year, and  
140 day of week was included as an indicator variable. Results of the first stage estimate the  
141 association between DTR and mortality for each city.

142

### 143 **2.3. Time varying distributed lag non-linear model**

144 The DLNMs, described in the first-stage analysis, assumed that the exposure-lag-response  
145 associations between DTR and mortality in each location were constant across the whole study  
146 period. We also applied a time-varying DLNM with a linear interaction (Gasparrini et al. 2015a;  
147 Gasparrini et al. 2016) between DTR and year. Using the time-varying DLNM, we derived  
148 coefficients representing the exposure-lag-response association for the first and last year of the  
149 study periods for each city. The set of four coefficients (the entire period, the first and the last  
150 year for each location) were reduced to one coefficient that modeled the overall cumulative  
151 associations between DTR and mortality. The sets of four coefficients were used to determine  
152 the lag-response relationships at the 99th percentile of DTR reference at 0°C DTR.

153

### 154 **2.4. Second stage meta-analysis**

155 We pooled one parameter of the overall cumulative exposure-response relationship and the four  
156 parameters of the lag-response relation. Multivariate random-effect meta-regression was used  
157 to pool the parameters by country. We used indicators of country as predictors in the meta-  
158 regression to country-pooled estimates and city-specific predicted parameters (Best Linear  
159 Unbiased Prediction, BLUP). Overall pooled coefficient (only for calculating excessive  
160 relative risk of overall countries) was estimated by meta-analysis without predictors. All  
161 analyses were performed using R software (version 3.3.1) packages `dlnm` and `mvmeta`  
162 (Gasparrini 2011; Gasparrini et al. 2012; Gasparrini et al. 2010).

163

## 164 **2.5. Attributable mortality risk fraction**

165 Overall cumulative relative risk estimated from BLUP for each city was used to compute the  
166 attributed number of deaths, and the fraction of deaths over the following 14 days at each  
167 location. **The total number of deaths attributed to DTR was calculated as the sum of all days in**  
168 **the series when DTR contributed to death and its ratio with the total number of deaths; this**  
169 **provides the ‘total attributable fraction’ (Gasparrini and Leone 2014).** We also computed the  
170 time-varying attributable risk of DTR based on BLUP for each city from the time-varying  
171 DLNM. Although time-dependent distributions of DTR and death could be used to estimate  
172 time-varying attributable risk, we used DTR and mortality distribution for the entire period  
173 because we did not find a clear difference between DTR distributions for the first and last three  
174 years of the series for each city (Table 1).

175

## 176 **2.6. Sensitivity analysis**

177 In order to test the sensitivity of our results to the modeling parameters and assumptions

178 described above, we changed lag days for DTR (21 days), inter-day temperature changes (10  
179 and 21 days), and the degrees of freedom (df) of lag knots for DTR (df=5), and analyzed the  
180 first results. We also assessed sensitivity to controlling for humidity (only for 6 countries which  
181 include relative humidity data), air pollution (Korea, O<sub>3</sub> and PM<sub>10</sub>), flexibility of long-term  
182 trend (df=7 and 9) and absolute temperature using various knot percentiles and changing lag  
183 days (14 and 28 days).

184

### 185 **3. Results**

186 Descriptive statistics of mortality, absolute temperature, and the distribution of DTR are in  
187 Table 1. Fig. 1 displays the geographical distributions of the 308 cities within the 10 countries  
188 included in the analyses and the corresponding annual averaged DTR ( $^{\circ}\text{C}$ ). The data set  
189 included 85,912,372 deaths. A variability in DTR was observed among countries over the entire  
190 study period, with mean values ranging from 6.7 $^{\circ}\text{C}$  (Ireland, 6 cities) to 10.9 $^{\circ}\text{C}$  (USA, 135  
191 cities). Table 1 also shows the DTRs and absolute temperature distributions during the first and  
192 second halves of the time periods for each country. As expected, the mean temperature  
193 increased slightly over time, although we did not detect a clear temporal pattern in the DTR  
194 values. City-specific descriptive statistics are reported in Supplementary Table S1.

195

196 The percent increases in risks and attributable mortality risk fraction of DTR estimated from  
197 the model with no interaction (i.e. the average throughout the study period) are reported in  
198 Table 2. Percent increases in risks of DTR (per 10 $^{\circ}\text{C}$ ) are highest in South Korea (6%, 95%  
199 CI: 3-9.1%), Spain (4.4%, 3-5.8%), and Brazil (4.2%, 1.7-6.7%). Colombia (-1.2%, -6.3-4.1%)  
200 and Ireland (0.3%, -3-3.8%) showed the lowest percent increases in risk of DTR, although

201 both were not significant. Table 2 also displays the total percentage of deaths attributable to  
202 DTR (reference at minimum DTR of each city, 2.5% with 95% empirical confidence interval  
203 (95% eCI): 2.3–2.7%). Similar with percent increases in risk, most of the country-specific  
204 estimated attributable risks were statistically significant. The risk fraction was highest in Korea  
205 (4.5%, 3-5.9%) and Spain (4.2%, 3.5-4.9%). The fractions were lowest in Colombia (-1.5%, -  
206 5.1–2.1%) and in Ireland (0.2%, -1.2–1.4%).

207

208 Fig. 2 displays the country-pooled lag-response associations at the 99th percentile of DTR  
209 referenced at 0°C. The coldest (Canada, Ireland, and UK) and warmest (Brazil, Colombia)  
210 countries showed the highest RR at lag 0 and lasted to a lag from 4–7 days. Other countries,  
211 which had moderate temperatures, have the highest RR in approximately 1–3 lag days and were  
212 limited to a lag of 7–14 days. The corresponding city-specific lag-response is displayed in  
213 Supplementary Fig. S1.

214

215 Results from an analysis of the temporal variation in the percent increases in risk of DTR are  
216 illustrated in Fig. 3. Table 3 displays temporal variation of estimates per year and test results  
217 for linear interaction (null hypothesis is the pooled interaction term is 0; the null hypothesis is  
218 that no temporal change occurred). The percent increases in risk of DTR increased from 2.5%  
219 (95% CI: 1.8–3.3%) to 3.8% (95% CI: 3.1–4.5%) between the first and last periods. Except for  
220 Ireland and Japan, all countries showed patterns of increasing percent increases in DTR risk,  
221 with -2.9–5% in the first year and 1.5–13.8% in the last year of the series. The temporal increase  
222 of percent increases in DTR risk were significant (P-value<0.05) in USA, UK, Spain, and South  
223 Korea (Table 3). Country-pooled temporal changes in the lag-response relationship are  
224 displayed in Supplementary Fig. S2. A comparison between the curves suggests that a longer

225 lag-association and smaller harvesting effect were observed in most countries.

226

227 Fig. 4 and Table 3 display the temporal variation in the attributable mortality risk fraction of  
228 DTR. In overall, the attributable risk fraction of deaths increased from 2.4% (95% eCI: 2.1-  
229 2.7%) to 2.7% (2.4-2.9%) between the first and last periods. The increase in the attributable  
230 risk fraction of death overtime was observed in all countries except Japan (0.07% decrease per  
231 a year) and Ireland (0.23% decrease per a year). Korea (0.56% per year) and Colombia (0.31%  
232 per year) showed the fastest increase of risk fraction, whereas Canada (0.03% per year), USA  
233 (0.09% per year) and UK (0.11% per year) showed the slowest increasing patterns.  
234 Corresponding city-specific estimates are reported in the Supplementary Material  
235 (Supplementary Table S2). And main conclusions were robust to sensitivity analysis  
236 (Supplementary Table S3).

237

#### 238 **4. Discussion**

239 Our findings show that DTR is responsible for a higher mortality risk increase (3.1%, 95% CI:  
240 2.7–3.5%) and fraction of deaths (2.5%, 95% eCI: 2.3–2.7%) in all the countries studied. South  
241 Korea and Spain showed the highest percent increase in risk (6% and 4.4%, respectively) and  
242 attributable risk fractions (4.5% and 4.2%, respectively). This study also provides evidence of  
243 the incremental health impact of DTR during the last few decades. With the exception of Japan  
244 and Ireland, an increasing pattern of percentile increases in risks (3.8% in the last year of the  
245 study periods, compared with 2.5% in the first year) was observed, and the attributable risk  
246 fraction showed the same temporal increasing pattern (2.7% in the last year of the study periods,  
247 compared with 2.4% in the first year).

248

249 This study is comparable to a recent temperature variability-mortality association study in the  
250 MCC Collaborative Network (Guo et al. 2016). Both studies were based on a similar multi-  
251 country data set and addressed the significant association between temperature variability and  
252 mortality, even after controlling for the main effect of absolute temperature. Guo et al.  
253 developed a new composite index of intra- and inter-day temperature variability using a  
254 standard deviation of minimum and maximum temperatures during the exposure days, and  
255 found the temperature variability-mortality relationship varied with exposure days (0–7 days),  
256 countries (twelve countries/regions with 372 communities), and season (cold, hot, and  
257 moderate). Meanwhile, our study only focused on the association between intra-day  
258 temperature variability and mortality, using a classical meteorology index (DTR) and flexible  
259 statistical method, which considers a flexible lag-response structure of DTR. In addition, our  
260 study included data from 308 cities in the 10 countries with more than 15 years of study data  
261 to estimate the temporal changes in the DTR-mortality association. It also described an overall  
262 increase in the health burden of DTR on mortality during recent decades.

263

264 Interestingly, our finding suggested that the DTR effects on mortality were higher in warm  
265 countries (Brazil, Australia, and Spain) compared to cold countries (Canada, Ireland, and UK),  
266 although Korea and Colombia were exceptions. These finding are consistent with previous  
267 studies, such as multi-country studies that reported that the effect of temperature variability  
268 with short exposure durations (0–1, and 0–2 days) on mortality is highest in hot area ( $>22.9^{\circ}\text{C}$ )  
269 than other areas (cold, moderate cold, and moderate hot areas) (Guo et al. 2016). U.S. studies  
270 also showed that higher DTR effect in southern areas (percent change of non-accidental  
271 mortality per one unit of DTR 0.24-0.31%) than other regions (0.22-0.27%) (Lim et al. 2014).

272 Studies in China also showed a similar trend of a relatively higher and more significant effect  
273 of DTR in warm cities (Guangzhou, and Shanghai) than cold cities (Anshan and Xi'an,  
274 although Tangshan is an exception). To specifically assess the association between the DTR  
275 effects and annual mean of absolute temperature, we fitted a weighted regression model  
276 (Supplementary Fig. S3); a city-specific BLUP of DTR coefficient (i.e. log of relative risk,  
277 estimated from the second stage analysis) was used as a response variable, annual mean  
278 temperature was used as an explanatory variable, and inversed city-specific variances of the  
279 DTR coefficient were used as weights. We observed a significantly positive linear association  
280 between the DTR effect and the annual mean temperature from the weighted regression model  
281 (P-value=0.01). This result can be interpreted as evidence to support the hypothesis that there  
282 may be an impact on mortality from the positive interactions between long-term temperature  
283 and DTR.

284

285 The synergism effect of the DTR and long-term temperature on mortality may be due to a  
286 number of factors. One mechanism may be aggravation. Hot temperature can disturb normal  
287 physiological thermoregulation, including changes in blood viscosity, plasma cholesterol level,  
288 and red blood cell count (Keatinge et al. 1986). Increasing DTR may also impact mortality  
289 through lowering the thermoregulatory system and negatively affecting the heart rate, heart rate  
290 variability, blood platelets, red blood cells, and blood viscosity (Keatinge et al. 1984b; Lim et  
291 al. 2012b). Because warm countries can be exposed to hot weather more often, the DTR effect  
292 of warm areas can be amplified by the increase in biological burden. Another hypothesis is that  
293 the effect of the DTR is higher in warm areas because people in warm and moderate areas are  
294 more likely to keep their windows open and spend more time outdoors, which may increase  
295 exposure to DTR, thus increasing the effect of DTR. However, our results only suggest the

296 possible associations of the aggravation hypothesis; further research should be conducted to  
297 find the causal relationship between DTR and long-term temperature on mortality.

298

299 Additional questions should be raised as to why the effect of DTR on mortality changed across  
300 time. We speculate several plausible explanations. The first hypothesis is deterioration by  
301 climate change, suggested in the previous paragraph. We found the higher risk and sharper risk  
302 increase in hot cities. **This finding suggests that climate change may increase the risk of DTR.**  
303 Secondly, an aging population may also be an important factor, as numerous studies have  
304 revealed that elderly people are more susceptible to DTR (Kan et al. 2007a; Lim et al. 2012a;  
305 Yang et al. 2013a), and the populations of developed and developing countries included in our  
306 research are aging (Börsch-Supan 2008; Faunce 2008). However, we could not identify the  
307 exact reason of the temporal increase, hence additional research regarding this topic is needed  
308 in further studies.

309

310 **Although not found in our study, prior studies have reported that climate change factors**  
311 **(greenhouse gases, urbanization, and aerosols) have led to global decline in the DTR during**  
312 **twenty century, because the nocturnal minimum temperatures have increased faster than**  
313 **maximum temperature (Braganza et al. 2004; Makowski et al. 2008). However, it is still unclear**  
314 **how this decline in the DTR will affect human health. Also, since the increasing nocturnal**  
315 **temperatures can affect mortality and distribution of DTR simultaneously, a confounding effect**  
316 **of the nocturnal temperatures needs to be considered to estimate the effect of DTR on death.**  
317 **Even if the effect of increasing nocturnal minimum temperature is partly considered in our**  
318 **study by controlling the daily averaged temperature, there is a limitation to control the effect**  
319 **of nocturnal temperature due to lack of data. Therefore, we expect more comprehensive studies**



320 to be carried out under various climate conditions with longer study periods and more detailed  
321 weather data.

322

323 As described earlier, although DTR and absolute temperature may affect human health in  
324 different ways, because both have a mechanism that negatively affects human health, the effects  
325 of absolute temperature and temperature variability on mortality have been an interesting topic  
326 of prior environmental research. In addition, comparing the health effects of two variables has  
327 important implications for understanding human health in a climate change context (which can  
328 increase both the average values and the variability of temperature)(Guo et al. 2016; Stocker  
329 2014; Vicedo-Cabrera et al. 2016). Recent studies asserted that DTR has a lesser effect than  
330 absolute temperature on mortality (Lee et al. 2017; Vicedo-Cabrera et al. 2016). Our results  
331 also suggest a lesser effect of DTR on mortality when compared with the total attributable  
332 mortality fraction of absolute temperature from a previous study (Gasparrini et al. 2015b). The  
333 total fraction of DTR attributed to mortality (2.5%) was much smaller than the fraction for total  
334 absolute temperature (7.71%)(Gasparrini et al. 2015b). However, our results may differ from  
335 the conclusions of previous studies (Chen et al. 2007; Kan et al. 2007a; Lim et al. 2015; Tam  
336 et al. 2009; Yang et al. 2013a) that used modeling strategies that did not fully control the flexible  
337 lag structures of absolute temperature. Because the effect of absolute temperature delayed up  
338 to several weeks of exposure, the estimates of DTR on mortality could be overestimated unless  
339 the main effects of temperature are fully adjusted. Hence, we contend that our modelling  
340 approach provides more appropriate results in estimating the health effects of DTR, in  
341 comparison with prior studies.

342

343 In this study, more acute DTR–mortality relationships (highest RR at 0 lag days) were observed

344 in warm and cold countries (Brazil, Colombia, Canada, Ireland, and UK). In contrast, more  
345 delayed DTR–mortality relationships (highest RR at 2–4 lag days) were observed in other  
346 moderate temperature countries. Although this study does not explain the difference, we  
347 speculate that the exact factors are related to the physiological, technological, and behavioral  
348 adaptations to the climate.

349

350 A key strength of our study is the use of a large multi-country, multi-city data set with different  
351 demographic distributions, climate conditions, and socio-economic characteristics. To the best  
352 of our knowledge, our study is the largest of its kind including 308 locations and more than 85  
353 million deaths from 10 countries. Our study also is the first and the largest study of time-  
354 varying DTR-related mortality, and the use of a uniform statistical framework across all cities  
355 makes the results directly comparable. In addition, unlike previous studies that have quantified  
356 the association on terms of RR (Kan et al. 2007b; Vicedo-Cabrera et al. 2016; Yang et al. 2013b),  
357 our study provides the attributable mortality burden of DTR. Because the attributable fraction  
358 considers the distribution and risks of each variable, we contend that the attributable fraction  
359 is a suitable measure to estimate mortality burden of the exposure variable and to establish  
360 corresponding public health strategies.

361

362 However, our study has some limitations that must be acknowledged. First, because regions of  
363 Africa, and large countries in Europe and Asia (such as France, Russia, and India) were not  
364 included in this study, our findings are not globally representative. Second, the data did not  
365 include age- or gender-specific mortality rates, which could be explored in future research.  
366 Third, we could only identify the suggestive association between DTR and all-causes mortality,  
367 and not the causal effect of DTR on mortality. Future studies should strive to overcome these

368 limitations by expanding the study populations and improving the study design.

369

## 370 **5. Conclusions**

371 In summary, this study finds that the significant DTR effect on mortality across all countries,  
372 and provides evidence that the effect of DTR was higher in warm regions. Although our  
373 estimated attributable mortality fraction of DTR is smaller than fractions of absolute  
374 temperature from a previous multi-country study (Gasparrini et al. 2015b), it is higher than  
375 fractions of extreme heat and cold temperature. In addition, although the risks and contributions  
376 of DTR on mortality varied for each country, it increased at the multi-country scale with  
377 significant increases estimated in USA, UK, Spain, and South Korea; and non-significant  
378 increments in Canada, Brazil, Colombia, and Australia. **The estimates of DTR-related mortality**  
379 **increased throughout the study period in overall regions, which could be interpreted as**  
380 **maladaptation to DTR. Consequently, there is a possibility that the health burden of DTR will**  
381 **increase in the near future.** Hence, we suggest that public-health policies and climate change  
382 research that have so far focused on the effects of extreme heat should be extended to account  
383 for the health burden of DTR and its temporal variation.

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512 **Table 1. Descriptive statistics by country.** Including distribution of diurnal temperature range  
513 in first 3 years (First) and last 3 years (Last) of country-specific study periods. USA: United  
514 States of America, UK: United Kingdom.

Country (# of city)	Time Period	Total Deaths	Study Period (year)	Absolute temperature (°C)	Diurnal temperature range (°C)					
				Mean	Mean	10%	25%	50%	75%	90%
<b>Canada</b> (26)	Whole	2,989,901	1986- 2011	6.8	10	4.4	6.6	9.7	13	15.9
	First			7	10.1	4.5	6.7	9.8	13	16
	Last			6.9	9.9	4.4	6.4	9.4	12.9	16.1
<b>USA</b> (135)	Whole	22,896,409	1985- 2006	14.8	10.9	5.6	7.8	10.6	13.9	16.7
	First			14.7	11	5.6	7.8	10.6	13.9	17.2
	Last			15.1	10.7	5.6	7.8	10.6	13.3	16.1
<b>Brazil</b> (18)	Whole	3,435,502	1997- 2011	24.2	9	5.4	6.8	8.6	10.7	13.2
	First			24.1	8.8	5.1	6.6	8.4	10.6	13.1
	Last			24.3	9.1	5.5	7	8.6	10.7	13.4
<b>Colombia</b> (5)	Whole	956,539	1998- 2013	23.4	9	5.8	7	8.8	10.8	12.4
	First			23.1	8.9	5.6	6.8	8.7	10.8	12.5
	Last			23.5	8.9	6.1	7.2	8.7	10.4	12.1
<b>UK</b> (10)	Whole	1,2075,786	1990- 2012	10.3	7.3	3.8	5.2	6.9	9.1	11.3
	First			10.1	7.3	3.8	5.1	6.9	9.1	11.3
	Last			10.1	7.5	3.9	5.3	7	9.4	11.7
<b>Ireland</b> (6)	Whole	1,058,215	1984- 2007	9.7	6.7	3.4	4.8	6.4	8.3	10.3
	First			8.9	6.8	3.6	4.9	6.5	8.4	10.4
	Last			10.3	6.9	3.6	4.9	6.6	8.5	10.5
<b>Spain</b> (51)	Whole	3,480,531	1990- 2010	15.5	10.6	4.9	7	10	13.8	17
	First			15.1	10.7	5	7.2	10.2	13.8	17.2
	Last			15.5	10.4	4.8	6.8	9.8	13.6	17
<b>Japan</b>	Whole	3,6113,897	1972- 2012	15.1	8.4	4.2	6	8.2	10.6	12.8

<b>(47)</b>	First			14.4	8.8	4.4	6.3	8.6	11	13.3
	Last			15.5	8.2	4.1	5.9	8	10.2	12.4
<b>South Korea</b>	Whole	1,727,642	1992-2010	13.7	8.2	4.1	5.9	8	10.2	12.7
<b>(7)</b>	First			13.5	8	3.8	5.6	7.7	10.1	12.5
	Last			13.8	8.3	4.3	5.9	8	10.3	12.7
<b>Australia</b>	Whole	1,177,950	1988-2009	18.1	8.2	4.4	5.9	7.8	10	12.4
<b>(3)</b>	First			18.1	7.8	4.1	5.6	7.3	9.5	11.9
	Last			18.7	8.1	4.5	5.8	7.6	10	12.6

516 **Table 2.** Percent increases in risk (per 10°C) and attributable risk fraction (%) of diurnal  
 517 temperature range on mortality by country. USA: United States of America, UK: United  
 518 Kingdom.

<b>Country</b>	<b>Percent Increases in Risk (%, 95% CI)</b>	<b>Attributable Risk Fraction (%, 95% eCI)</b>
Canada	2.6 % ( 0.9 , 4.2 )	2.7 % ( 1.8 , 3.5 )
USA	2.9 % ( 2.3 , 3.6 )	3.2 % ( 2.9 , 3.5 )
Brazil	4.2 % ( 1.7 , 6.7 )	3.7 % ( 2.6 , 4.9 )
Colombia	-1.2 % ( -6.3 , 4.1 )	-1.5 % ( -5.1 , 2.1 )
UK	2.9 % ( 1.5 , 4.4 )	2.1 % ( 1.6 , 2.7 )
Ireland	0.3 % ( -3 , 3.8 )	0.2 % ( -1.2 , 1.4 )
Spain	4.4 % ( 3 , 5.8 )	4.2 % ( 3.5 , 4.9 )
Japan	3.1 % ( 2.3 , 3.9 )	2.7 % ( 2.4 , 3 )
South Korea	6 % ( 3 , 9.1 )	4.5 % ( 3 , 5.9 )
Australia	4.2 % ( 0.7 , 7.9 )	3.3 % ( 1.1 , 5.3 )
Overall	3.1 % ( 2.7 , 3.5 )	2.5 % ( 2.3 , 2.7 )

519 eCI: empirical confidence interval.

520 **Table 3.** Variation of excessive relative risk (per 10°C) and attributable fraction (%) of diurnal  
 521 temperature range on mortality per a year, and p-value of the test. USA: United States of  
 522 America, UK: United Kingdom.

Variation (per year)				
Country	Study Period (Years)	Percent Increases in Risk	Attributable Risk Fraction	p-value*
Canada	26	0.06 %	0.03 %	0.4384
USA	22	0.09 %	0.09 %	0.0209
Brazil	15	0.25 %	0.23 %	0.2243
Colombia	16	0.29 %	0.31 %	0.6583
UK	23	0.17 %	0.11 %	0.0344
Ireland	24	-0.35 %	-0.23 %	0.0989
Spain	21	0.28 %	0.23 %	0.0025
Japan	41	-0.09 %	-0.07 %	<0.0001
South Korea	19	0.78 %	0.56 %	<0.0001
Australia	22	0.3 %	0.26 %	0.1543

523 \* Significant test on temporal change by Wald type test of the pooled reduced coefficient of the  
 524 year-interaction terms. The null hypothesis is that no change in year occurred.

525 **Figure legends**

526 **Fig. 1.** Geographical locations of study cities and their annual mean values of diurnal  
527 temperature range (DTR, °C).

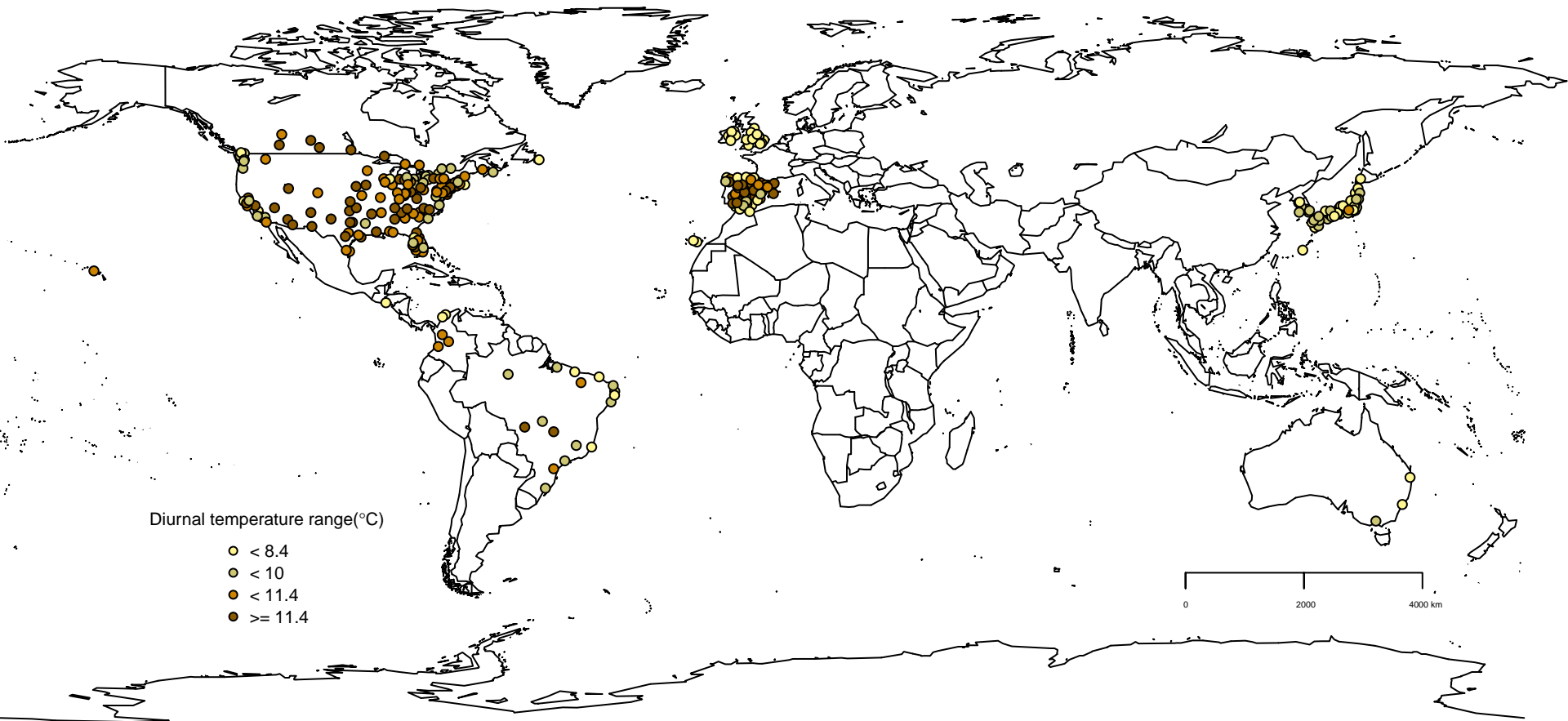
528 **Fig. 2.** Lag-response relationship between diurnal temperature range (DTR) and mortality  
529 predicted for the overall study periods of 10 countries; RR: relative risk. USA: United States,  
530 UK: United Kingdom.

531 **Fig. 3:** Temporal changes in percent increases in risk (%) between the first (First) and the last  
532 (Last) year of country-specific study periods; USA: United States, UK: United Kingdom.

533 **Fig. 4.** Temporal changes in attributable risk fraction (%) between the first (First) and the last  
534 (Last) year of country-specific study periods; USA: United States, UK: United Kingdom.

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537 **Competing financial interest:** The authors declare they have no actual or potential competing  
538 financial interests.

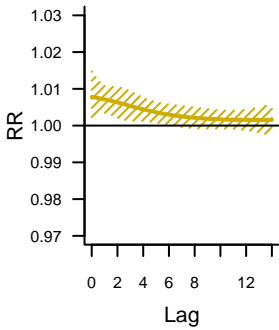
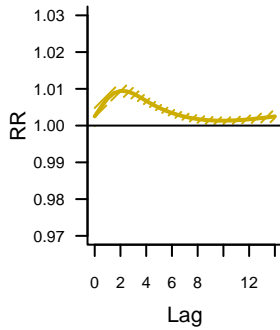
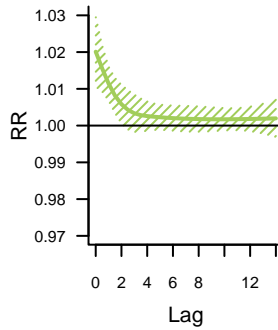
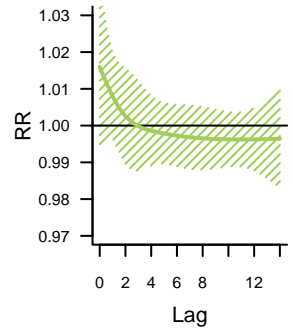
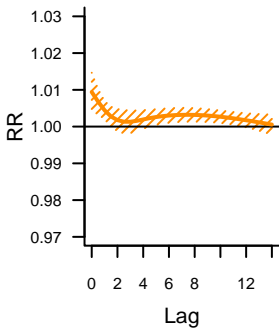
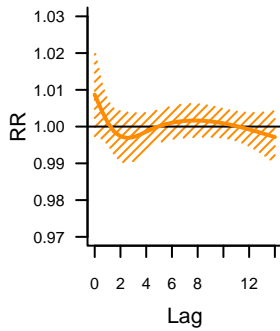
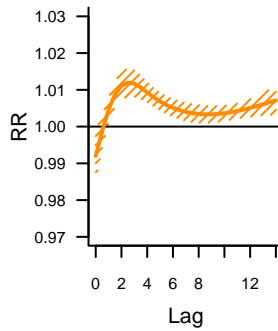
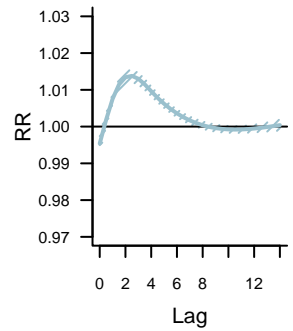
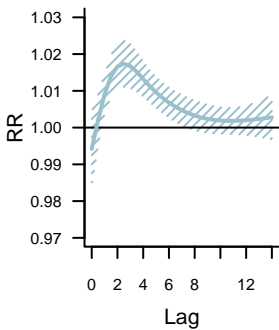


Diurnal temperature range(°C)

- < 8.4
- < 10
- < 11.4
- $\geq 11.4$

0 2000 4000 km



**Canada****USA****Brazil****Colombia****UK****Ireland****Spain****Japan****South Korea****Australia**