

1 **Disproportionate climate burden of rising temperature on low birth**
2 **weights in a health-compromised nation**

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23

24 **Summary**

25 **Background:** Warming temperatures add to the global health burden, with profound implications for
26 vulnerable populations including pregnant women and newborns. Low birth weight is a major neonatal
27 health issue in Pakistan leading to neonatal mortality and impaired long-term health. This study assessed
28 the impact of extreme temperatures on low birth weights, developed heat vulnerability index at district-
29 level, identified high-risk subgroups, estimated heat-attributable burden, and projected future risks.

30 **Methods:** We constructed a space-time series study using national survey data from 2008–2017 across
31 Pakistan. Distributed-lag nonlinear models within a generalised linear mixed-effects framework
32 accounted for provincial variation. We modelled nonlinear and lagged associations between temperature
33 and low birth weights using quadratic functions, and model-averaged predictions addressed uncertainty.
34 Province-level risk estimates combined with district-level indicators, such as live births, child mortality,
35 poverty index, and heat index, developed the heat vulnerability index.

36 **Findings:** The study included 85,017 participants, with 15,920 (18·72%) infants identified as having low
37 birth weights. Heat-related risks for low birth weights varied across provinces, with relative risks ranging
38 from 1·30 (1·07–1·57, 95% confidence interval) to 1·74 (1·16–2·59). The heat-related population
39 attributable fraction ranged from 8·95 to 10·91%, translating to 1·12 million heat-related low birth-weight
40 cases over the study period. Projections estimate that heat-related population attributable fractions will
41 increase by 1·23–2·36% by the 2060s. Subgroup analysis showed that the risk of heat-related low birth
42 weight was associated with hazardous air pollution, women with both lower or higher education, and
43 those living in urban areas. Women in southern Punjab, northern Baluchistan, and Sindh faced the highest
44 risks.

45 **Interpretation:** Our findings identify Pakistan's districts most vulnerable to heat-related low birth weight
46 and highlight contributing factors. These insights can drive targeted interventions to mitigate risks. The
47 study advances understanding of global impacts of rising temperatures, particularly in resource-limited
48 and high-risk settings.

49

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51 not-for-profit sectors.

52

53 **Keywords:** child health, climate change, extreme heat, global health, heat-related population attributable
54 fraction, heatwaves, infants, pregnant women, relative risk

55

56

57 **Introduction**

58 Rising temperatures and the increased frequency of extreme heat due to climate change present a major
59 threat to global health. These impacts are more pronounced in low- and middle-income and resource-
60 limited settings. In South Asia, heatwaves are now 45 times more likely than in the pre-industrial era,
61 with temperatures 0.85°C higher due to climate change, making it one of the most vulnerable regions
62 globally (1). Pakistan, already facing socio-economic and health challenges, is among the countries most
63 affected by climate change (2). The country has experienced an increase in extreme weather events —
64 heatwaves, floods, and droughts — attributed to shifting climate patterns.

65 Extreme heat disproportionately affects marginalised populations, including pregnant women who are
66 more at risk due to poverty, malnutrition, limited healthcare access, and resource scarcity (3). With a
67 neonatal mortality rate of 41 deaths per 1,000 live births and 19–32% of infants born low birth weight,
68 Pakistan bears one of the highest burdens (4). Low birth weight is a leading cause of neonatal mortality
69 and is linked to long-term health impairments such as stunted growth and cognitive deficits (4).

70 Pregnant women are particularly vulnerable to heat-related complications due to physiological
71 susceptibility and limited access to healthcare (5). Seasonal variation in birth outcomes suggests that
72 temperature extremes affect the maternal biological and behavioural responses. Research on temperature
73 effects on birth outcomes is abundant in high-income countries with robust birth registries, but such
74 studies are scarce in low-resource settings like Pakistan, where populations are more vulnerable to
75 climate-related impacts.

76 In addition to heat, worsening air pollution in Pakistan's urban centres has emerged as a large health
77 concern, especially for pregnant women and newborns (6). As of 2023, Pakistan ranks as the fourth most-
78 polluted country globally, with air pollution reducing life expectancy by 3·9 years (7). Although air
79 pollution is not the primary focus of our study, we accounted for it as a confounding factor due to its
80 impact on maternal and neonatal health.

81 Pakistan presents a good opportunity to investigate heat impacts on resource-limited and marginalised
82 communities. The nation is highly susceptible to climate change impacts and faces compounded
83 challenges including growing population, limited healthcare infrastructure, widespread poverty, and
84 socio-political instability. These overlapping vulnerabilities highlight the urgency of localised evidence to
85 inform targeted interventions and policies. Our study provides comprehensive evidence on heat exposure,
86 socio-economic disparities, and maternal health in Pakistan, offering insights to mitigate climate-related
87 health inequities both nationally and in other low-resource settings.

88 Our objective is to assess extreme weather's impact on low birth weight in Pakistan, identify high-risk
89 areas and subgroups, and project future risks under two climate-change scenarios. We aims to provide
90 temperature-specific risk estimates at a provincial scale and develop a district-level heat vulnerability

91 index to guide targeted interventions and policies. By integrating maternal health and environmental
92 datasets within a robust statistical framework, we generate insights into climate variability and low birth
93 weight in Pakistan, with implications in other low- and middle-income countries.

94

95 **Research in context**

96 **Evidence before this study**

97 The body of research on the impacts of rising temperatures and heat exposure on health outcomes,
98 particularly maternal and child health, is expanding rapidly. Current evidence strongly suggests that the
99 risk of low birth weight is associated with heat exposure. However, most of this evidence originates from
100 high-income countries, while only a few recent studies have explored the impacts of heat on birth weights
101 in resource-limited and vulnerable settings. In particular, Pakistan experiences major challenges in
102 maternal and child health, ranking among the worst globally for these indicators. Pakistan is also one of
103 the most severely affected by climate change, experiencing frequent heatwaves, floods, and droughts.
104 Despite this vulnerability, only two studies to date have examined the relationship between heat exposure
105 and low birth weight in Pakistan — one focusing on a single district and one providing national-scale
106 estimates through global analyses. There is therefore a major research gap at both national and local
107 scales to inform targeted intervention strategies and identify high-risk factors.

108

109 **Added value of this study**

110 Our study starts to fill this gap by providing both national and regional evidence for the association
111 between heat exposure and low birth weight in Pakistan. This is a challenging endeavour where
112 comprehensive national datasets are often unavailable. While studies in low- and middle-income
113 countries frequently rely on Demographic and Health Surveys and Multiple Indicator Cluster Surveys,
114 these datasets have limitations, including inconsistent data collection, missing data, and until recently, the
115 absence of some data such as the exact day of birth. Neither are spatial coordinates provided for most
116 datasets from Pakistan. Despite these constraints, our study used robust methodologies to produce
117 provincial-level, long-term risk estimates for the relationship between heat and low birth weights. We also
118 downscaled these estimates to develop a district-level heat vulnerability index for low birth weights,
119 enabling the identification of high-risk areas.

120 In addition to leveraging Demographic and Health Surveys and Multiple Indicator Cluster Surveys, we
121 integrated multiple global and national datasets, including air pollution indices, poverty indices, and
122 mortality rates, to account for a wide range of factors influencing low birth weights. Furthermore, we
123 provided projections of future risks for low birth weights under different climate-change scenarios,

124 offering valuable insights for policymakers and public health stakeholders to design targeted interventions
125 that enhance resilience in vulnerable regions.

126

127 **Implications of all the available evidence**

128 This study provides context-specific evidence from Pakistan, identifying the country's vulnerability to
129 climate change and warming temperatures, thereby enabling the design and implementation of targeted
130 interventions. Pakistan's maternal and child health outcomes are poorer than those in countries of similar
131 socio-economic status and challenges. For example, Pakistan's maternal mortality rate is 319 per 100,000
132 live births, compared to 124 in similar countries; the neonatal mortality rate is 41 per 1,000 live births,
133 compared to 20 per 1,000 elsewhere. Limited healthcare access and widespread socioeconomic
134 inequalities are among the largest contributors to these disparities.

135 By introducing interventions in high-risk areas with economic constraints, improvements in maternal
136 and child health outcomes can be achieved. This goal aligns directly with the country's Sustainable
137 Development Goals (SDGs), particularly SDG 3 (Good Health and Well-being), SDG 10 (Reduced
138 Inequalities), and SDG 13 (Climate Action).

139 Pakistan's National Action Plan for climate change is hindered by economic constraints, inadequate
140 institutional coordination, and challenges in resource mobilisation. In such a resource-constrained
141 environment, studies like ours can play a pivotal role in guiding the development of cost-effective,
142 targeted interventions. For example, leveraging successful existing programs, such as the Lady Health
143 Workers program, could provide an efficient platform for delivering necessary health and climate-related
144 information to women in vulnerable regions. The development of a district-level heat vulnerability index
145 provides a replicable framework for identifying high-risk areas and designing targeted interventions,
146 which can be applied in other low- and middle-income countries experiencing climate-related health
147 challenges.

148 **Methods**

149 **Data Sources**

150 ***Health and socio-economic data***

151 We compiled data from 11 surveys done between 2010–2020, including Pakistan Demographic Health
152 Surveys (DHS) (8) and UNICEF Multi-Indicator Cluster Surveys (MICS) (9), covering maternal and
153 child health, socio-economic status, and demographics. We defined low birth weight as ≤ 2.5 kg or based
154 on maternal recall of the baby being “smaller than average.” While recall is subjective and prone to bias,
155 there is 86% agreement with measured birth weights (10). We did sensitivity analyses to assess
156 associations of heat with birth weights and birth sizes individually. Data sources, cleaning, and processing
157 are detailed in Supplementary Table S1.

158

159 ***Environmental data***

160 We obtained monthly gridded meteorological data (2008–2017) from Copernicus ERA5-Land (9×9 km
161 resolution), including mean temperature ($^{\circ}\text{C}$), precipitation (m), and dew point ($^{\circ}\text{C}$) (11). We estimated
162 the heat index using the weathermetrics R package (12). We sourced mean monthly $\text{PM}_{2.5}$
163 concentrations ($\mu\text{g m}^{-3}$) from the Atmospheric Composition Analysis Group (11-km resolution) (13). We
164 aggregated district-level environmental variables to the provincial level using arithmetic means.

165

166 **Study design**

167 ***Space-time series design***

168 We applied a space-time series design to analyse long-term, delayed, and non-linear associations between
169 monthly mean temperatures and low birth weights at the provincial level. We addressed spatial non-
170 independence using generalised linear mixed-effects models with random intercepts and captured
171 temporal delays in temperature effects through a distributed-lag non-linear framework (14). We
172 calculated weighted averages across models to derive coefficients and variance matrices. This approach
173 effectively resolved uncertainties, data sparsity, sampling gaps, and potential missingness inherent in
174 survey-based datasets.

175

176 ***Data aggregation***

177 We linked monthly health and environmental data by province, month, and year. We analysed the data at
178 the provincial level to address zero inflation in district-level data and capture regional trends. We
179 aggregated low birth weight cases and environmental variables monthly, using temporal indicators
180 (month and year of birth) to compensate for the absence of exact birth dates in the survey data.

181

182 **Analysis**

183 ***Modelling framework***

184 We modelled the association between monthly mean temperature exposure and low birth weights using
185 generalised linear mixed-effects models with a negative binomial error distribution to account for
186 overdispersion. We included random intercepts to account for provinces-level non-independence. We
187 modelled the exposure-response relationship using linear and quadratic functions (polynomial degree = 2)
188 and specified a lag of seven months (polynomial degree = 1) to capture long-term lagged effects during
189 pregnancy. The selection of the lag period was informed by existing studies (15) and further validated in
190 sensitivity analysis.

191 We adjusted for PM_{2.5}, maternal education, wealth index, humidity and precipitation in six different
192 models. We compared models based on Akaike's information criterion (AIC), with model weights used to
193 predict model-averaged risks at specific percentiles. We imputed missing data for education and wealth
194 index using the mice package in R (16). Table S2 and Fig. S1, Supplementary Materials present details
195 on model comparisons and handling of missing data.

196

197 ***Subgroup analyses***

198 We did subgroup analyses based on: **maternal education**: low (no education or primary) *versus* high
199 (secondary or higher), **wealth index**: low (poorest, poorer, and middle) *versus* high (richer and richest),
200 **air quality**: PM_{2.5} < 25 µg m⁻³ ('fair') *versus* PM_{2.5} ≥ 25 µg m⁻³ ('poor/hazardous'), and **region type**:
201 urban *versus* rural. We estimated subgroup-specific risks using model-averaged coefficients and
202 crossbasis functions and reported results for specific temperature percentiles.

203

204 ***Heat vulnerability index***

205 We developed a district-level heat vulnerability index to identify high-risk regions by incorporating
206 indicators for exposure, sensitivity, and adaptive capacity (17). We assessed environmental exposure
207 using district-level average heat index (derived from mean temperature and humidity) and PM_{2.5}
208 concentration over the study period. We represented sensitivity with under-five mortality rates and total
209 live births, while the multidimensional poverty index captured adaptive capacity, reflecting socio-
210 economic vulnerabilities. We integrated estimates of province-specific relative risk to establish empirical
211 relationships between heat exposure and health outcomes. To ensure comparability, we scaled and centred
212 variables (mean = 0, SD = 1). Supplementary Materials, Fig. S3, provide further details on index
213 construction, scaling methods, and uncertainty modelling.

214

215

216 ***Temperature-related population attributable fractions***

217 We estimated the population attributable fraction of low birth-weight cases attributable to temperature
218 exposure under current and future climate scenarios. For the baseline (2008–2017), we used observed
219 mean temperature data from ERA5-Land. For future projections (2048–2057 and 2068–2077), we applied
220 projected mean temperature increases obtained from the World Bank's Climate Change Knowledge Portal
221 (climateknowledgeportal.worldbank.org). The scenarios included Shared Socio-Economic Pathways
222 (SSP) SSP2-4.5 ('intermediate' greenhouse-gas emissions) and SSP5-8.5 ('very high' greenhouse-gas
223 emissions) (18).

224 For each scenario, we calculated the median temperature over the entire study period to represent
225 population acclimatisation. We defined heat exposure as temperatures above the median, and cold
226 exposure as temperatures below the median. We calculated population attributable fractions as:

227
$$100 \frac{r_{\text{dlm}} - 1}{r_{\text{dlm}}} P \quad [\text{eq 1}]$$

228 where r_{dlm} = mean relative risk derived from distributed lag nonlinear models associated with exposure
229 over the study period, and P = proportion of days classified as heat or cold exposure. We applied Monte
230 Carlo simulations (1,000 iterations) to incorporate uncertainty in risk estimates. We calculated the
231 confidence limits for the 2·5th and 97·5th percentiles of the simulated population attributable fractions.

232

233 ***Sensitivity analysis***

234 We did sensitivity analysis based on birth weights and birth size, different lag periods, and different
235 exposure and lag degrees. Further details on sensitivity analysis are provided in Supplementary Materials
236 (Table S5).

237

238 ***Software and tools***

239 We did all data preprocessing and analyses in R (version 4.1.0; R Core Team 2024) with the `dlnm` (19),
240 `glmmTMB` (20), and `MuMIn` (21) packages, and visualisation with `ggplot2` (22). All code and data are
241 available at doi:10.5281/zenodo.14373384.

242

243 ***Role of funding source***

244 No funding source.

245 **Results**

246 After cleaning and preprocessing the data, we retained 85,017 records from mothers who provided birth
247 weight or size information for their most recent child born within the last five years during the study
248 period. Among these, 15,920 children had a low birth weight (≤ 2.5 kg) or were reported smaller than
249 average in size at birth. By extrapolating to the national level and using an average birth rate of 29.69 per
250 1,000 population (from 2008 to 2017) and annual total population estimates (23), we calculated that
251 approximately 60.15 million babies were born in Pakistan between 2008 and 2017 (mean = 6.01 million
252 per year), of which 18.72% (~ 11.26 million) had low birth weight.

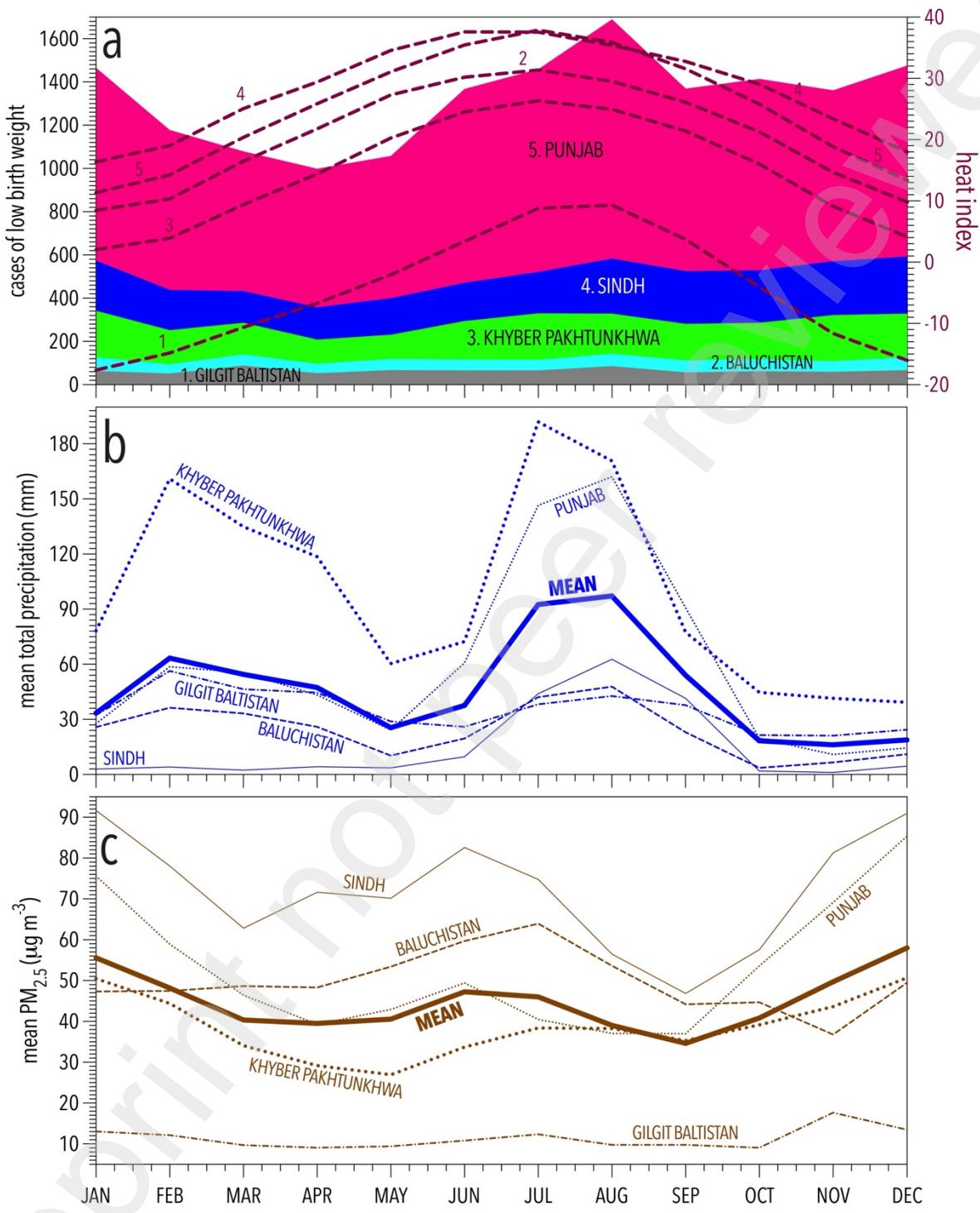
253 Nearly three-quarters (74.65%) of participants had primary or no education, 13.47% had completed
254 secondary education, and 11.87% had attained higher education (Table 1). About one-third (29.01%) of
255 participants belonged to the ‘poorer/poorest’ wealth category, 12.57% were in the middle-income group,
256 and 23.79% were classified as ‘richer/richest’ (Table 1). The median temperature during the study period
257 was 19.5 °C, and the median air pollution concentration was 44.8 $\mu\text{g m}^{-3}$ (Table 1).

258

259 -----INSERT TABLE 1 HERE----

260

261 The graphical analysis of monthly trends for low birth weight, mean temperature, precipitation, humidity,
262 and air pollution ($\text{PM}_{2.5}$) revealed that cases of low birth weight are higher during hotter months when
263 temperatures, humidity, and precipitation are highest (Fig. 1).

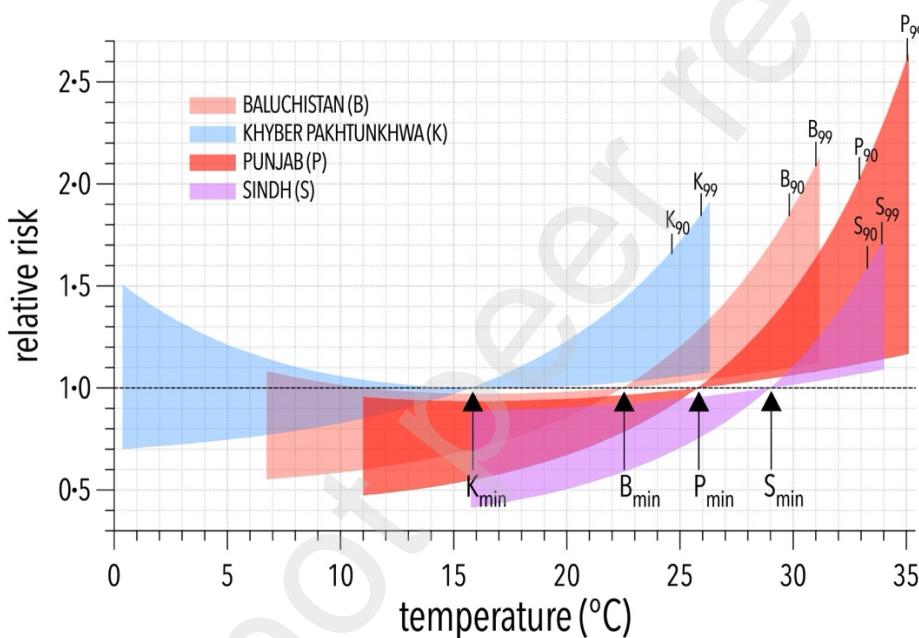


264

265 **Figure 1.** Monthly trends of (a) cases of low birth weight along with average heat index,
266 average precipitation,
267 and (c) average particulate matter (PM_{2.5}) concentration.

268 ***Province-level exposure-response association between temperature and low birth weights***

269 We observed a strong positive association between low birth weight and heat exposure across all
270 provinces (Fig. 2). The relative risk estimates range from 1·30 (1·07–1·57, 95% confidence bounds) to
271 1·48 (1·11–1·97) at moderate heat (90th percentile), and from 1·37 (1·09–1·73) to 1·74 (1·16–2·59) at
272 extreme heat (99th percentile), compared to the median reference temperature (Fig. 2). In contrast, we
273 found no evidence of a relationship between the risk of low birth weight and cold exposure for all
274 provinces except Gilgit Baltistan. In our findings, heat exposure had an impact; therefore, the results and
275 discussions primarily emphasise the effects of heat. We present province-specific risk estimates for all
276 provinces in Table S3 Supplementary materials.



277

278 **Figure 2.** Exposure-response association between counts of low birth weight and mean temperature (with 95%
279 confidence bounds). Arrows indicate the minimum risk, while province-level letters (B = Baluchistan, K = Khyber
280 Pakhtunkhwa, P = Punjab, S = Sindh) and vertical lines represent the 90th, and 99th percentiles for relative risk
281 estimation.

282

283 Our stratified subgroup analysis indicates that higher air pollution exacerbates the risk of low birth
284 weight during hot conditions 2·11 (1·23–3·64). Women with primary or no education had a 1·99 (1·20–
285 3·30) relative risk of delivering low birth weight infants during extreme heat. Women with secondary or
286 higher education also had an increased risk of low birth weight associated with extreme weather exposure
287 (2·36; 1·18–4·74) (Table 2).

288 Similarly, women from both high and low wealth groups had a high risk of low birth weight linked to
289 extreme heat 2·03 (1·21–3·41) and 7·06 (1·68–29·68), respectively. The results with wide confidence

290 intervals should be interpreted with caution. There was also an elevated heat-related risk among women in
291 urban (2·22; 1·24–3·96) *versus* rural areas (2.02; 1·21–3·37) (Table 2).

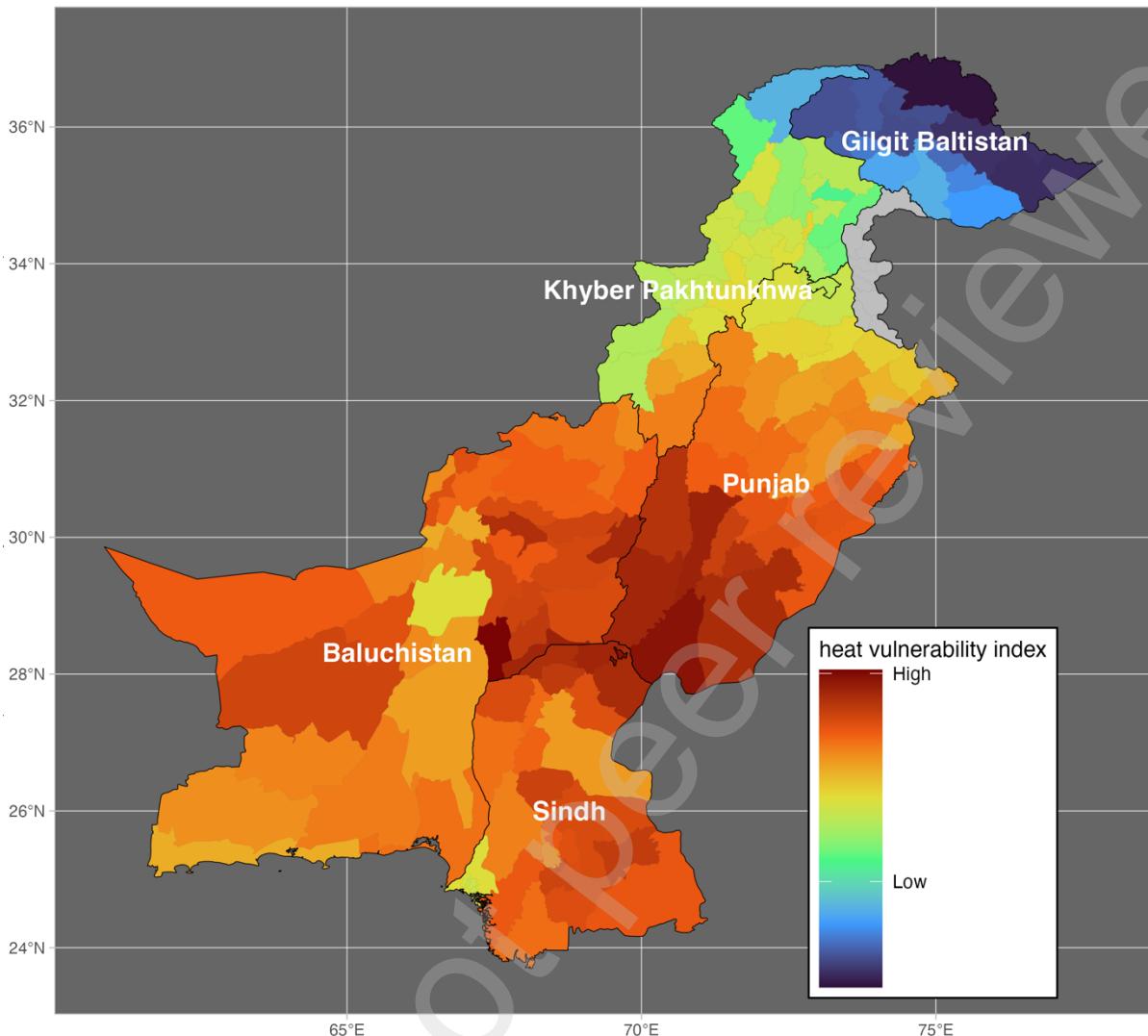
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293 -----INSERT TABLE 2 HERE----

294

295 ***District-level heat vulnerability index***

296 The heat vulnerability index map (Fig. 3) illustrates the spatial distribution of heat vulnerability for low
297 birth weights across Pakistan's provinces and districts. The map highlights that districts in southern
298 Punjab, northern Baluchistan, and Sindh exhibit the highest vulnerability, indicating a greater risk of heat-
299 related low birth weights in these regions. These areas are likely influenced by higher heat exposure and
300 compounding socio-economic and environmental conditions such as poverty and air pollution. In
301 contrast, northern regions such as Gilgit Baltistan and parts of Khyber Pakhtunkhwa had lower
302 vulnerability due to their cooler climates.



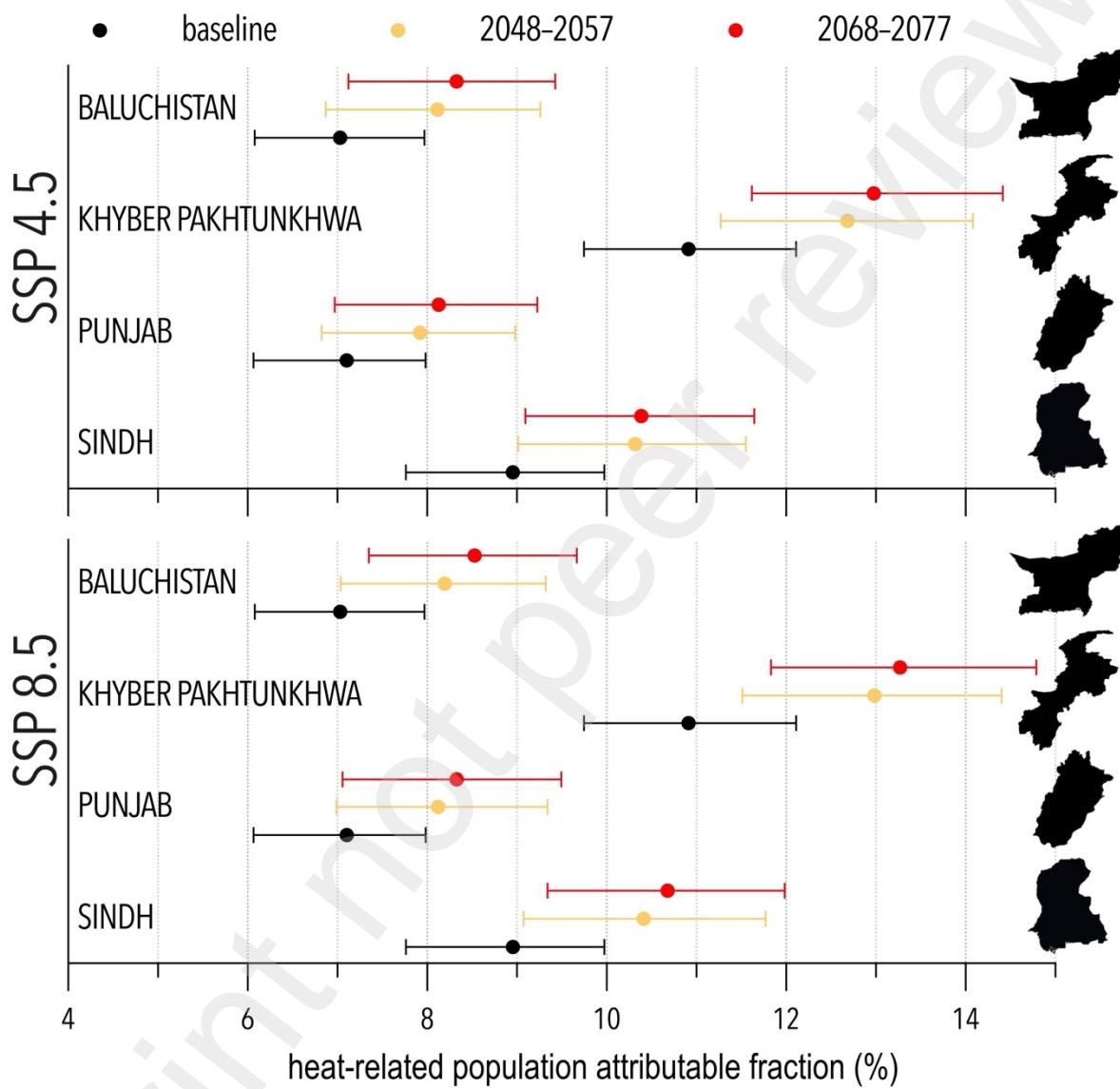
303
304 **Figure 3.** District-level spatial patterns of the heat vulnerability index of low birth weights associated with heat.
305

306 ***Current and projected population attributable fraction of heat-related low birth weights***

307 The analysis of the population attributable fraction revealed regional variation in temperature-related
308 impacts on low birth weight under current and future climate scenarios (Fig. 4; Table S4). In Baluchistan,
309 the heat-related population attributable fraction increased from 8·95% (7·76–9·97) during the baseline
310 period to 10·68% (9·35–11·98) under SSP5-8·5 for 2068–2077, representing an increase of 1·73% by the
311 2060s under a high-emissions scenario. Similarly, Khyber Pakhtunkhwa experienced an increase in the
312 heat-related population attributable fraction from 7·10% (6·07–7·98) at baseline to 8·33% (7·06–9·50)
313 under SSP5-8·5 for 2068–2077, showing an increment of 1·23% by the 2060s.

314 Punjab had the highest increase in the heat-related population attributable fraction, rising from 10·91%
315 (9·75–12·11) at baseline to 13·27 (11·83–14·79) under SSP5-8·5 for 2068–2077, with an increase of
316 2·36% by the 2060s. In Sindh, the heat-related population attributable fraction increased from 7·03%

317 (6.08–7.97) at baseline to 8.53% (7.35–9.67) under SSP5-8.5 for 2068–2077 (increasing by 1.50% by
 318 the 2060s). These findings underscore the intensifying impacts of heat under high-emissions scenarios,
 319 highlighting regional disparities in vulnerability to temperature-related risks on low birth weight.



320
 321 **Figure 4.** Heat-related population attributable fraction (%) with 95% confidence intervals by province for climate-
 322 projection scenarios SSP 4.5 (upper panel) and SSP 8.5 (lower panel).

323 **Discussion**

324 Our study highlights the impact of heat exposure on low birth weight in Pakistan, emphasising the health
325 implications of rising temperatures and compounded vulnerabilities in resource-limited settings. Pakistan
326 stands out for having some of the worst maternal and child-health outcomes among low- and middle-
327 income countries with similar socio-economic status (24). Climate change and extreme temperature
328 conditions further exacerbate this vulnerability. Dimitrova et al. (25) highlighted that Pakistan (along with
329 Mali, Sierra Leone, and Nigeria) experiences a disproportionately high burden of neonatal mortality
330 linked to temperature fluctuations, with > 160 temperature-related neonatal deaths per 100,000 live births.
331 Similarly, Zu et al. (26) estimated the temperature-related burden of low birth weight across several
332 countries, reporting the greatest reduction in birth weight — -257g (-498·06 to -17·15g) — among
333 children in Pakistan.

334 These statistics underscore the intersection of climate vulnerability and systemic healthcare challenges
335 in Pakistan, highlighting that heat exposure is not just an environmental issue, but also a social
336 determinant of health that disproportionately impacts marginalised populations. Our study builds on this
337 evidence, providing detailed insights into the risk of low birth weights associated with heat (30–70%),
338 identifying contributing factors, pinpointing high-risk areas, and estimating future population attributable
339 fractions. This localised analysis informs targeted interventions and adaptation strategies for regions
340 facing severe climate-related health challenges.

341 We found that women with both primary and secondary/higher education are vulnerable to the effects
342 of extreme heat (27). Similarly, women from both higher and lower-income households have a high risk
343 of low birth weights. This counterintuitive vulnerability could be attributed to several factors, including a
344 lack of adaptation to extreme heat in urbanised environments where the urban heat-island effect is
345 pronounced. Frequent power outages are common in Pakistan and can worsen the effects of heat
346 exposure, even in households equipped with air conditioning (28). In wealthier households, medical
347 interventions such as elective C-sections and other obstetric procedures, might increase physical stress in
348 pregnant women, making them more susceptible to heat-related complications (29). Sedentary lifestyles,
349 often associated with wealthier populations, might further impair thermoregulation, making it more
350 difficult to cope with extreme temperatures (27).

351 We identified that Punjab was the most vulnerable to heat-related low birth weights. The heat
352 vulnerability index revealed that districts in lower Punjab, southern Sindh, and northern Baluchistan are
353 the most at risk. These regions are among the hottest in Pakistan and face substantial risks from climate
354 change. The spatial analysis underscores the importance of prioritising these areas for public health
355 interventions, providing local health authorities with scientific evidence to inform resource allocation and
356 health policies. We estimated that 8·95 to 10·91% of low birth-weight cases in Pakistan were attributable

357 to hot conditions across various provinces. Based on this estimate, if approximately 11·26 million
358 children were born with low birth weight in Pakistan from 2008 to 2017, an average of 1·12 million cases
359 can be attributed to hot weather conditions. These estimates are projected to increase by 1·23–2·36% in
360 high-emissions scenarios by the 2060s. The estimates suggest that high temperatures are a major threat to
361 progress in maternal health and could widen social and health inequalities in low- and middle-income
362 countries.

363 We did not observe an association between cold exposure and low birth weight, except for Gilgit-
364 Baltistan, a region with a harsh tundra climate and severe winters. The area's high altitude, extreme cold,
365 and limited healthcare access during winter likely contribute to this finding. Further research is needed to
366 understand the combined impacts of heat, glacier melting, flooding, and landslides in these terrains and
367 their interaction with maternal and child health to guide adaptation strategies.

368 Pakistan's vulnerability is intensified by its rapidly growing population, young average age, and
369 cultural barriers limiting women's autonomy and access to prenatal care. Political instability, civil unrest,
370 and periodic terrorism further strain the country's capacity to prioritise public health. Limited healthcare
371 infrastructure and research resources hinder the development of adaptive strategies to address climate-
372 related health impacts. Together, these factors create a backdrop of socio-economic fragility, heightening
373 the vulnerability of pregnant women and newborns to adverse health outcomes, including low birth
374 weight. Addressing the intersection of climate change and maternal health requires comprehensive
375 strategies that integrate climate adaptation and mitigation with strengthening health systems, promoting
376 gender equality, and building community resilience. Collaborative efforts at the global, regional, and
377 national scales are essential to mitigate the impact of climate change on maternal health and to achieve
378 sustainable development goals.

379 Educational programs should inform pregnant women about heat risks and mitigation strategies, while
380 the government must improve access to prenatal check-ups and maternal healthcare services. Investments
381 in prenatal and postnatal care in high-risk areas are necessary for enhancing maternal and child health.
382 Tailored early warning systems can alert pregnant women and healthcare providers to heatwaves through
383 mobile networks, local media, and antenatal care centres. Lady Health Workers and non-governmental
384 organisations should promote heat adaptation, hydration, and support access to cooling facilities.
385 Collaborations with the Pakistan Meteorological Department and telecommunication companies can
386 extend outreach via mobile apps, voice messages, and community networks.

387 As climate change progresses, it will be necessary to develop and test adaptation strategies to
388 understand and improve the vulnerabilities of pregnant women in urban and rural areas to heat-related
389 health risks. Continuous monitoring and evaluation of these strategies will be essential to ensure their
390 effectiveness over time. With the heat vulnerability maps developed, high risk areas can be specifically

391 targeted for interventions, optimising the allocation of resources and improving outcomes for the most
392 vulnerable populations.

393

394 ***Limitations***

395 We used data from Pakistan's Demographic and Health Surveys and Multiple Indicator Cluster Surveys,
396 representative of national maternal and child health indicators. However, these datasets have limitations:
397 (1) The lack of exact childbirth dates required monthly data aggregation, which might de-emphasise
398 short-term temperature peaks, even though the approach effectively captures broader temporal patterns.
399 Aggregated data also enable reliable long-term burden estimation, aligning with the principles of
400 Basagaña and Ballester (30). (2) Missing spatial coordinate required aggregation at monthly and
401 provincial scales, accounting for regional variation while addressing data sparsity. To complement this,
402 we developed a district-level heat vulnerability index to identify high-risk areas for targeted interventions.
403 (3) We addressed inherent variability and gaps in survey data using weighted coefficients from multiple
404 models and random intercepts for provinces. (4) Birth size, a crude measure used alongside birth weight,
405 can introduce biases. (5) We excluded data from the 2017–2018 Pakistan Demographic and Health
406 Survey to avoid redundancy and overlap with the 2018 Multi-Indicator Cluster Surveys. (6) Gridded
407 meteorological data served as a proxy for exposure, capturing local trends but lacking individual-level
408 spatial and temporal precision. Monthly averages might obscure variability and household factors
409 influencing low birth weight.

410 The study's aggregated nature limits control over individual-level confounders, so we encourage
411 caution when interpreting the relationships observed as causal. Including spatial coordinates and precise
412 childbirth dates in Pakistan's Multi-Indicator Cluster Surveys, along with robust maternal and child health
413 data systems, are essential. Improved data collection will enhance understanding of environmental
414 impacts on health and support more targeted interventions.

415

416 **Conclusions**

417 Pregnant women in Pakistan face a heightened risk of delivering infants of low birth weight following
418 exposure to extreme temperatures, with percent risk varying among provinces from 30 to 70%. Around
419 8·95 to 10·91% of cases of low birth weights are attributable to heat and are projected to increase by 1·2
420 to 2·36% for climate projections to the 2060s. Regions in southern Punjab, northern Sindh, and
421 Baluchistan have the highest risks.

422 Addressing this issue requires the development, implementation, and rigorous evaluation of
423 comprehensive intervention strategies. These approaches must be multifaceted, leveraging collaboration
424 between researchers, government agencies, and local communities. Enhancing and adapting existing

425 strategies will be essential for improving health outcomes and protecting maternal and child health in the
426 future.

427

428 **Contributors**

429 SHF: Conceptualisation, Data Curation, Formal Analysis, Methodology, Visualisation, Project
430 Administration, Writing-original draft preparation. CJAB: Formal Analysis, Methodology, Visualisation,
431 Writing-reviewing and editing. ZAB: Validation, Writing-reviewing and editing. PB: Validation, Writing-
432 reviewing and editing. JKD: Validation, Writing-reviewing and editing. SM: Conceptualisation,
433 Validation, Writing-reviewing and editing. ZSL: Conceptualisation, Validation, Funding Acquisition,
434 Writing-reviewing and editing

435

436 **Declaration of interests**

437 We declare no competing interests.

438

439 **Data Sharing**

440 All relevant R code and data for the analyses and results available at doi:10.5281/zenodo.14373384.

441

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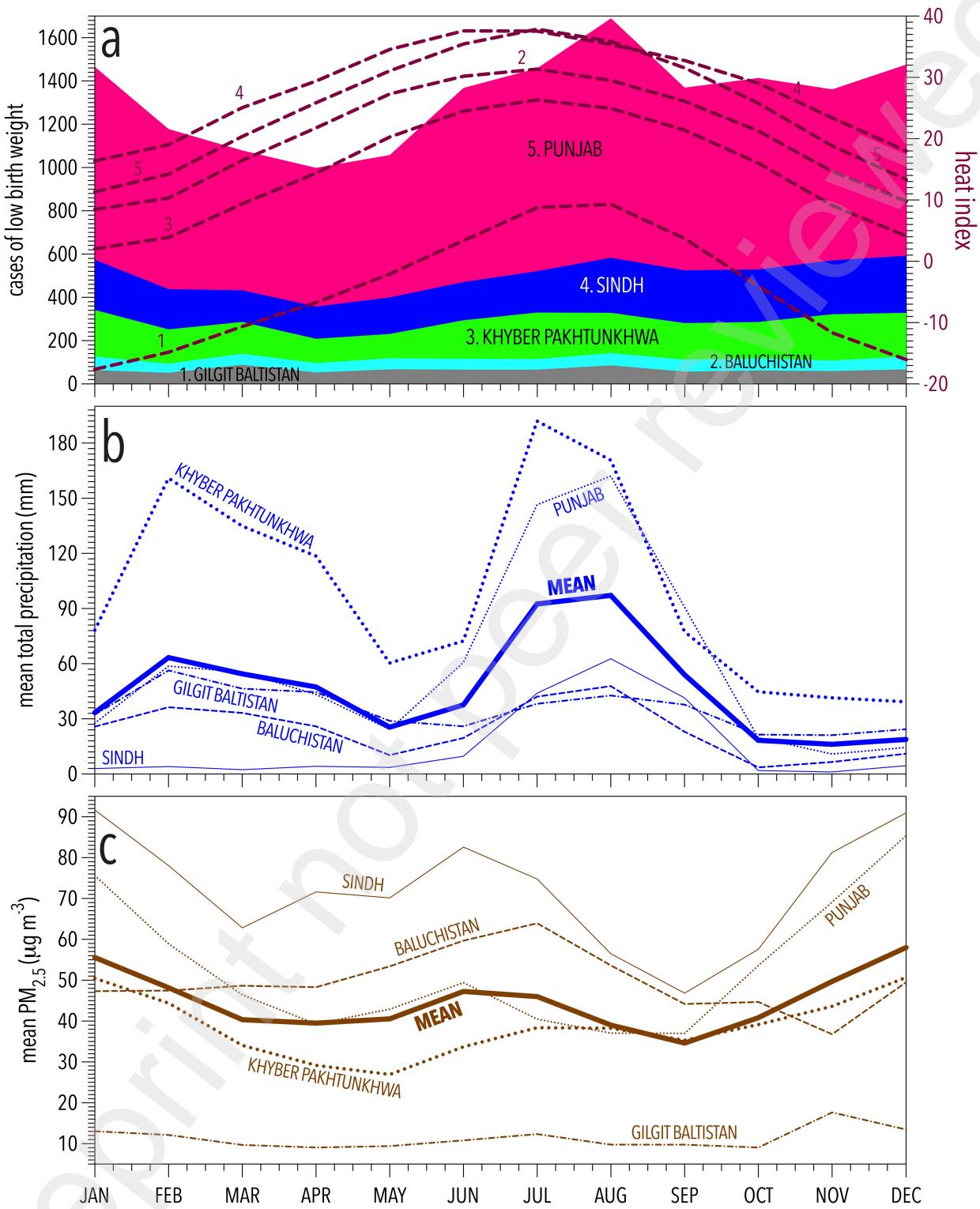
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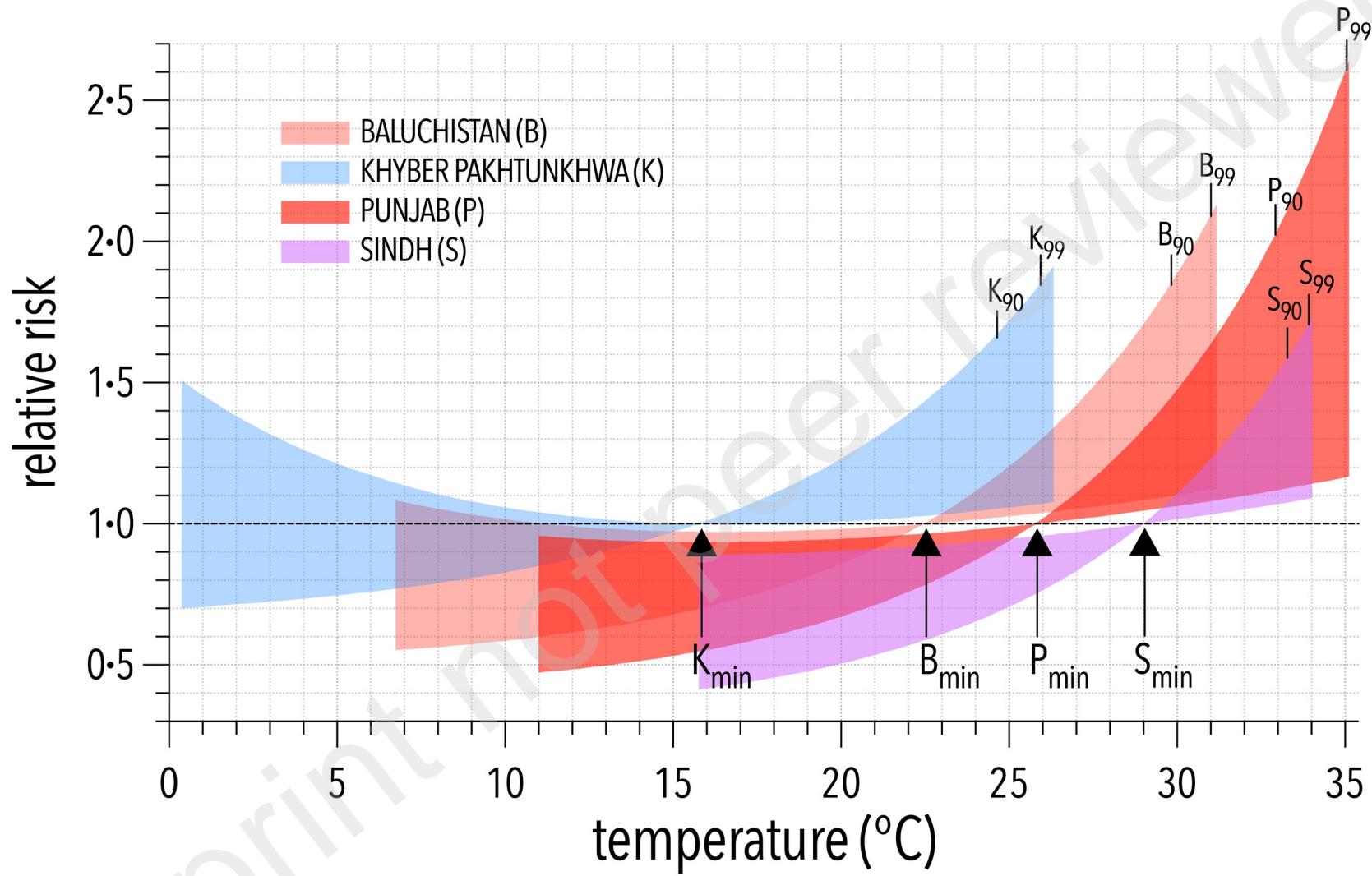
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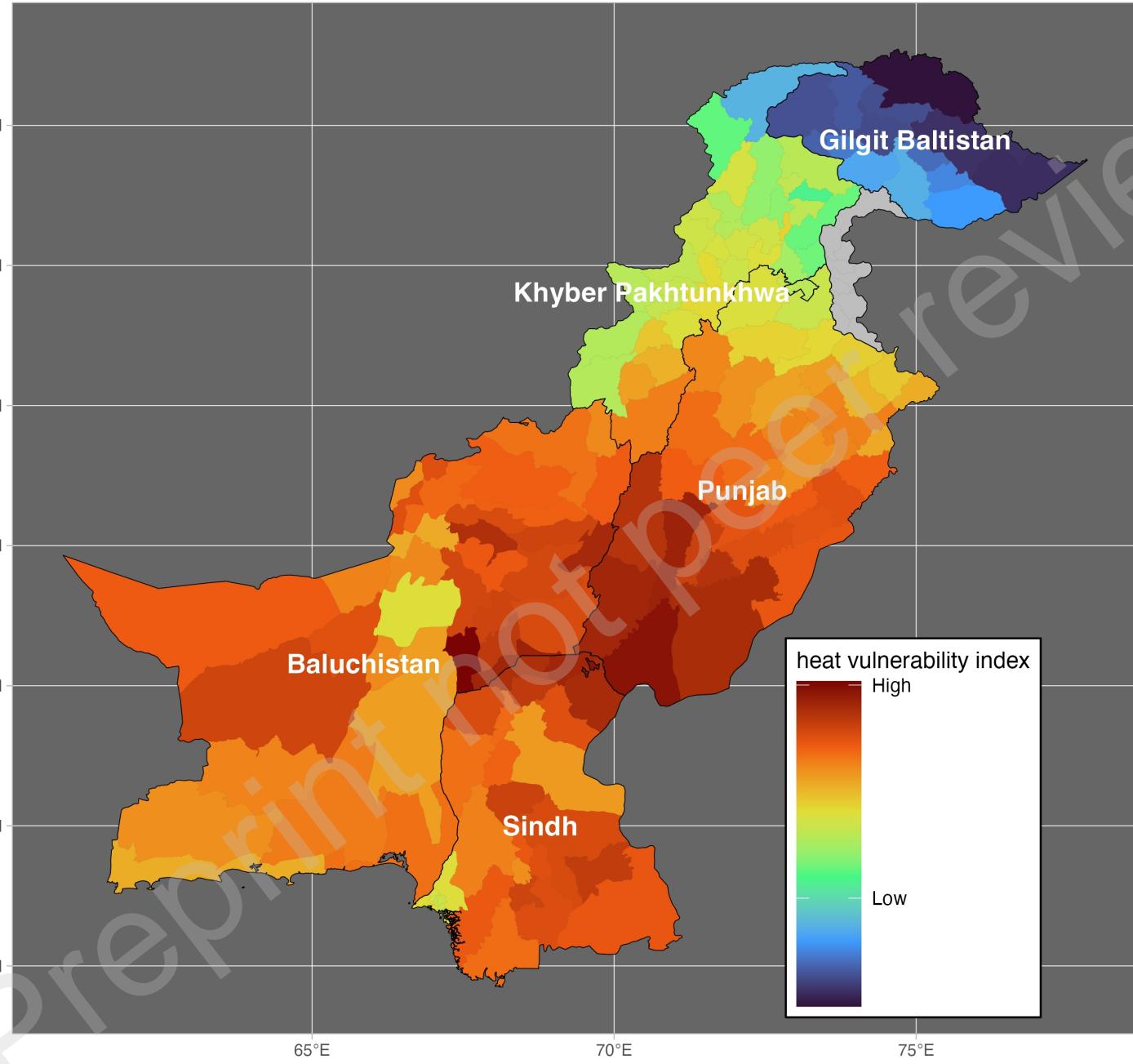
446 **References**

- 447 1. World Weather Attribution. Climate change made the deadly heatwaves that hit millions of highly
448 vulnerable people across Asia more frequent and extreme. 2024. Available from:
449 <https://www.worldweatherattribution.org/climate-change-made-the-deadly-heatwaves-that-hit-millions-of-highly-vulnerable-people-across-asia-more-frequent-and-extreme>.
- 450 2. Eckstein D, Künzel V, Schäfer L. Global Climate Risk Index 2021: Who suffers most from extreme
451 weather events? Weather-related loss events in 2019 and 2000-2019. Germanwatch. 2021. Available from:
452 <https://www.germanwatch.org/en/cris>.
- 453 3. Islam M, Ali S, Majeed H, et al. Drivers of stunting and wasting across serial cross-sectional household
454 surveys of children under 2 years of age in Pakistan: potential contribution of ecological factors. Am J Clin
455 Nutr. 2025;35:S0002-9165(25)00003-6.
456 <https://doi.org/10.1016/j.ajcnut.2025.01.003>
- 457 4. Nisar YB, Dibley MJ. Determinants of neonatal mortality in Pakistan: secondary analysis of Pakistan
458 Demographic and Health Survey 2006–07. BMC Public Health. 2014;14:1-12.
459 <https://doi.org/10.1186/1471-2458-14-663>
- 460 5. Shankar K, Hwang K, Westcott JL, et al. Associations between ambient temperature and pregnancy
461 outcomes from three South Asian sites of the Global Network Maternal Newborn Health Registry: a
462 retrospective cohort study. BJOG. 2023;130:124-33.
463 <https://doi.org/10.1111/1471-0528.17616>
- 464 6. Anwar A, Ayub M, Khan N, Flahault A. Nexus between air pollution and neonatal deaths: a case of Asian
465 countries. Int J Environ Res Public Health. 2019;16(21):4148.
466 <https://doi.org/10.3390/ijerph16214148>
- 467 7. Lee K, Greenstone M. Annual update: Air Quality Life Index. 2021. Available from:
468 https://aqli.epic.uchicago.edu/wp-content/uploads/2021/08/AQLI_2021-Report.EnglishGlobal.pdf
- 469 8. Demographic Health Program. DHS overview. The DHS Program website. Funded by USAID. [Internet].
470 [cited 2024 Feb 9].
471 Available from: <http://www.dhsprogram.com>.
- 472 9. UNICEF. Multiple indicator cluster surveys. [Internet]. 2024 [cited 2024 Feb 9].
473 Available from: <https://www.mics.unicef.org/surveys>.
- 474 10. Acharya P, Adhikari S, Adhikari TB. Mother's perception of size at birth is a weak predictor of low birth
475 weight: evidence from Nepal Demographic and Health Survey. PLoS One. 2023;18(1):e0280788.
476 <https://doi.org/10.1371/journal.pone.0280788>
- 477 11. Muñoz-Sabater J, Dutra E, Agustí-Panareda A, et al. ERA5-Land: a state-of-the-art global reanalysis
478 dataset for land applications. Earth Syst Sci Data. 2021;13(9):4349-83.
479 <https://doi.org/10.5194/essd-13-4349-2021>
- 480 12. Anderson GB, Bell ML, Peng RD. Methods to calculate the heat index as an exposure metric in
481 environmental health research. Environ Health Perspect. 2013;121(10):1111-9.
482 <https://doi.org/10.1289/ehp.1206273>
- 483 13. Van Donkelaar A, Hammer MS, Bindle L, et al. Monthly global estimates of fine particulate matter and
484 their uncertainty. Environ Sci Technol. 2021;55(22):15287-300.
485 <https://doi.org/10.1021/acs.est.1c05309>
- 486 14. Gasparrini A, Armstrong B, Kenward MG. Distributed lag non-linear models. Stat Med. 2010;29(21):2224-
487 34. <https://doi.org/10.1002/sim.3940>
- 488 15. Nyadanu SD, Tessema GA, Mullins B, Chai K, Yitshak-Sade M, Pereira G. Critical windows of maternal
489 exposure to biothermal stress and birth weight for gestational age in Western Australia. Environ Health
490 Perspect. 2023;131(12):127017. <https://doi.org/10.1289/ehp12660>
- 491 16. van Buuren S, Groothuis-Oudshoorn K. mice: multivariate imputation by chained equations in R. J Stat
492 Softw. 2011;45(1):1-67. doi:10.18637/jss.v045.i03. <https://doi.org/10.18637/jss.v045.i03>

- 494 17. Bao J, Li X, Yu C. The construction and validation of the heat vulnerability index: a review. Int J Environ
495 Res Public Health. 2015;12(7):7220-34. <http://doi.org/10.3390/ijerph120707220>
- 496 18. Intergovernmental Panel on Climate Change (IPCC). Summary for policymakers. In: Climate Change
497 2021: The Physical Science Basis. Cambridge University Press; 2021. p. 3-32.
<https://doi.org/10.1017/9781009157896.001>
- 498 19. Gasparrini A. Distributed lag linear and non-linear models in R: the package dlnm. J Stat Softw.
499 2011;43(8):1. <https://doi.org/10.18637/jss.v043.i08>
- 500 20. Brooks ME, Kristensen K, van Benhem KJ, et al. glmmTMB balances speed and flexibility among
501 packages for zero-inflated generalized linear mixed modeling. R J. 2017;9(2):378-400.
<https://doi.org/10.32614/RJ-2017-066>
- 502 21. Bartoń K. MuMIn: Multi-model inference. R package version 1.48.4. CRAN.R-
503 project.org/package=MuMIn; 2024. Available from: [https://cran.r-
504 project.org/web/packages/MuMIn/MuMIn.pdf](https://cran.r-project.org/web/packages/MuMIn/MuMIn.pdf)
- 505 22. Wickham H. ggplot2: Elegant graphics for data analysis. Springer-Verlag; 2016. Available from:
<https://cran.r-project.org/web/packages/MuMIn/MuMIn.pdf>
- 506 23. United Nations Children's Fund (UNICEF). Maternal and newborn health disparities in Pakistan. 2024.
507 Available from: [https://data.unicef.org/wp-
508 content/uploads/country_profiles/Pakistan/country%20profile_PAK.pdf](https://data.unicef.org/wp-content/uploads/country_profiles/Pakistan/country%20profile_PAK.pdf).
- 509 24. Aziz A, Saleem S, Nolen TL, et al. Why are the Pakistani maternal, fetal and newborn outcomes so poor
510 compared to other low and middle-income countries? Reprod Health. 2020;17:1-12.
<https://doi.org/10.1186/s12978-020-01023-5>
- 511 25. Dimitrova A, Mengel M, Gasparrini A, Lotze-Campen H, Gabrysch S. Temperature-related neonatal deaths
512 attributable to climate change in 29 low-and middle-income countries. Nat Commun. 2024;15(1):5504.
<https://doi.org/10.1038/s41467-024-49890-x>
- 513 26. Zhu Z, Zhang T, Benmarhnia T, et al. Estimating the burden of temperature-related low birthweight
514 attributable to anthropogenic climate change in low-income and middle-income countries: a retrospective,
515 multicentre, epidemiological study. Lancet Planet Health. 2024;8(12):e997-1009.
[https://doi.org/10.1016/S2542-5196\(24\)00242-0](https://doi.org/10.1016/S2542-5196(24)00242-0)
- 516 27. Conte Keivabu R, Cozzani M. Extreme heat, birth outcomes, and socioeconomic heterogeneity.
517 Demography. 2022;59(5):1631-54. <https://doi.org/10.1215/00703370-1017483>
- 518 28. ABC News. South Asian heatwave: Heatwave hits India and Pakistan as power stations struggle to secure
519 coal supply. 2022 Apr 30 [cited 2025 Jan 17]. Available from: [https://www.abc.net.au/news/2022-04-
520 30/south-asian-heatwave/101027906](https://www.abc.net.au/news/2022-04-30/south-asian-heatwave/101027906)
- 521 29. Mumtaz S, Bahk J, Khang YH. Rising trends and inequalities in cesarean section rates in Pakistan:
522 Evidence from Pakistan Demographic and Health Surveys, 1990-2013. PLoS One. 2017;12(10):e0186563.
<https://doi.org/10.1371/journal.pone.0186563>
- 523 30. Basagaña X, Ballester J. Unbiased temperature-related mortality estimates using weekly and monthly
524 health data: a new method for environmental epidemiology and climate impact studies. Lancet Planet
525 Health. 2024;8(10):e766-77. [https://doi.org/10.1016/S2542-5196\(24\)00212-2](https://doi.org/10.1016/S2542-5196(24)00212-2)







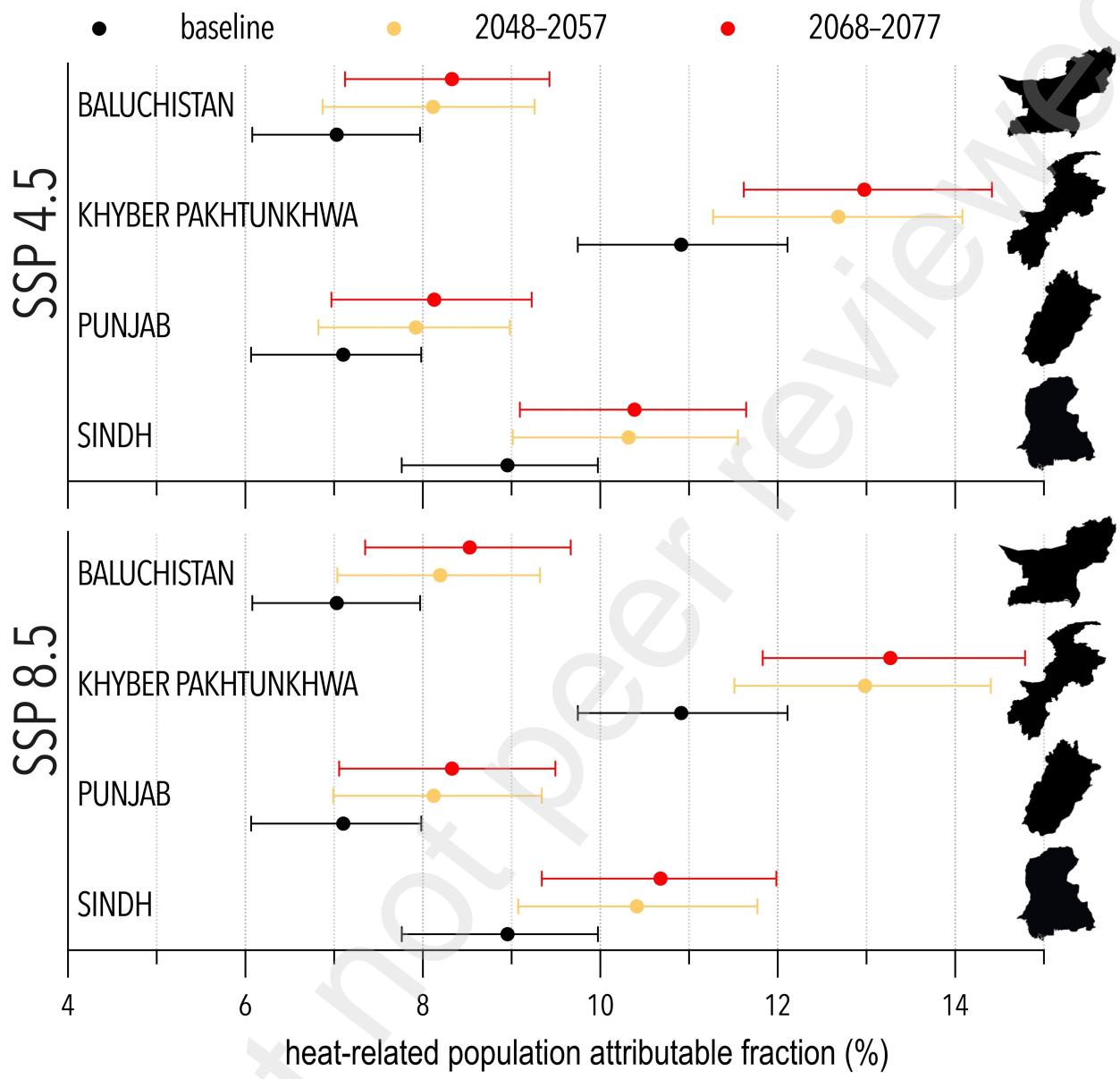


Table 1. Prevalence of low birth weight, socio-demographic status, and environmental conditions in Pakistan from 2008 to 2017 ($n = 85,017$).

Demographic and socioeconomic variables			
Variable	%	variable	%
prevalence / 100 cases ($n = 15,920$)	18·72		
education		mother Age	
primary or no education (%)	74·65	15–29	59·37
secondary (%)	13·47	30–39	35·53
higher (%)	11·87	40–49	5·11
wealth index		Province	
poor	29·01	Baluchistan	3·95
middle	12·57	Gilgit Baltistan	4·98
rich	23·79	Khyber Pakhtunkhwa (including FATA)	13·02
region		Punjab	62·31
rural	66·69	Sindh	15·74
urban	33·31		
Environmental variables			
variables	median (range)	variable	median (range)
mean temperature ($^{\circ}\text{C}$)	19·5 (-20·0–35·1)	humidity (%)	54·3 (18·9–80·2)
air pollution PM _{2·5} ($\mu\text{g m}^{-3}$)	44·8 (4·5–123·9)	precipitation (mm)	46·60 (0·03–427)

Table 2. Relative risk estimates (with 95% confidence bounds) of low birth weight linked to heat and cold across heat-index percentile ranges.

Subgroup	risk estimates (90th Percentile)	risk estimates (99th Percentile)
air quality status		
fair ($< 25 \mu\text{g m}^{-3}$)	0·80 (0·51–1·27)	0·98 (0·53–1·80)
poor–hazardous ($\geq 25 \mu\text{g m}^{-3}$)	1·78 (1·17–2·71)	2·11 (1·23–3·64)
education Status		
primary or no education	1·70 (1·15–2·50)	1·99 (1·20–3·30)
secondary or above	1·69 (1·04–2·76)	2·36 (1·18–4·74)
wealth Status		
low	1·72 (1·15–2·58)	2·03 (1·21–3·41)
high	4·54 (1·50–13·71)	7·06 (1·68–29·63)
type of region		
urban	1·94 (1·20–3·14)	2·22 (1·24–3·96)
rural	1·74 (1·16–2·62)	2·02 (1·21–3·37)