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1 Incidence and Burden of Cardiovascular Disease Attributable to

2 Extreme Heat in China

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3

4 **Abstract**

5 **Aim** The increasing frequency of heat events driven by climate change poses a serious
6 challenge to cardiovascular health. This study aimed to investigate the relationship between
7 high temperature and acute cardiovascular disease (CVD) incidence in China, and to
8 quantify the heat-related burden of CVD.

9 **Methods** 856,357 incident acute CVD cases were used from CVD surveillance in China
10 in 2023. A distributed lag non-linear model was applied to evaluate the dose-response
11 relationship between temperature and CVD incidence, as well as the best linear unbiased
12 prediction of the minimum incidence temperature (MIT). The heat-related burden was
13 quantified using attributable fraction and attributable number for two temperature ranges:
14 all heat and extreme heat.

15 **Results** Higher temperatures were associated with an increased risk of acute CVD
16 incidence. Extreme heat exposure resulted in a cumulative relative risk (RR) of 1.17 (95%
17 confidence interval[CI] 1.05-1.30), with the highest cumulative risk observed on the fourth
18 day following exposure. All heat exposure accounted for 3.19% of CVD cases, while
19 extreme heat contributed to 0.08%. The burden was particularly pronounced among
20 individuals aged ≥ 65 years (RR = 1.20, 95% CI 1.05-1.37), rural populations (RR = 1.18,
21 95% CI 1.01-1.37), and those living in temperate monsoon and temperate continental
22 climates (RR = 1.25, 95% CI 1.05-1.49). MIT varied geographically, ranging from 16.0 °C
23 in northern regions to 26.2 °C in southern regions, with the highest MITs concentrated in
24 tropical areas.

25 **Conclusions** These findings emphasize the urgent need for region-specific public health
26 strategies that integrate climate change adaptation and CVD prevention to mitigate the
27 growing health risks associated with rising temperatures.

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2 **Lay summary**

3 This study examined the relationship between high temperatures and acute cardiovascular
4 disease (CVD) incidence in China and quantified the heat-related burden.

- 5 ● Higher temperatures increased the risk of acute CVD, with extreme heat exposure
6 leading to a 17% higher risk, peaking on the fourth day after exposure.
- 7 ● The heat-related burden was more pronounced among older adults (≥ 65 years), rural
8 populations, and residents of the northern region, highlighting the need for targeted
9 public health strategies.

10 **Keywords:** Attributable fraction, Cardiovascular disease, Environmental exposure,
11 Extreme heat, Incidence and burden

12

13 **Introduction**

14 Significant public health achievements in cardiovascular disease (CVD) prevention have
15 been achieved in recent decades, largely attributed to interventions targeting traditional risk
16 factors such as high blood pressure, elevated cholesterol, and smoking ¹⁻³. Despite this
17 progress, the burden of CVD remains substantial, particularly with the increasing trends of
18 urbanization and aging populations. In 2022, CVD was responsible for an estimated 19.8
19 million deaths worldwide, resulting in 396 million years of life lost (YLLs) and an

1 additional 44.9 million years lived with disability (YLDs)⁴. Entering the era of precision
2 medicine, further reductions of the CVD burden require expanding prevention strategies
3 beyond conventional risk factors to incorporate nontraditional ones, such as environmental
4 exposures, which can provide novel insights into CVD prevention. Due to increased
5 greenhouse gas emissions from human activities, global warming has become increasingly
6 severe. This lead to a significant rise in both the frequency and the intensity of extreme
7 heat⁵, imposing a substantial burden on cardiovascular health.

8 High temperature can lead to physiological changes, including increased heart rate,
9 heightened blood viscosity, and dehydration due to excessive sweating, which collectively
10 place additional strain on the heart and increase the risk of CVD⁶. Numerous studies have
11 documented heat-related CVD mortality, consistently showing trends that align with
12 established physiological mechanisms⁷⁻¹⁰. With the shift in CVD prevention strategies
13 toward early health interventions, it becomes even more critical to explore the relationship
14 between high temperature and CVD incidence, and the burden on the healthcare system
15 should also be measured in terms of incidence. However, relatively few studies have
16 investigated the relationship between high temperature and the risk of incident CVD¹¹⁻¹³,
17 and even fewer have quantified heat-related incidence burden in terms of attributable
18 fraction (AF) or attributable number (AN)¹⁴, particularly across multiple counties in China.
19 Furthermore, the effect of ambient temperature on CVD hospitalizations varies by
20 population characteristics¹². This highlights the importance to better understand heat-
21 related risks in the context of CVD and underscores the need for targeted, population-
22 specific prevention strategies.

23 To address these gaps, we use the most up-to-date and comprehensive registration data to
24 investigate the relationship between heat exposure and acute CVD incidence in China.
25 Moreover, we calculate the AF and AN, which serve as burden measures that extend

beyond RR by estimating the actual impact of heat exposure on population health. These measures are critical for informing targeted interventions and policies to ultimately strengthen resilience against the health impacts of climate change.

Methods

Data sources

This study was based on the CVD surveillance in China, a nationwide surveillance launched in 2021. This system covered approximately 238 million people, accounting for 16.9% of the population of China, and aimed to assess the incidence of acute CVD among Chinese residents aged 18 years and older. All medical institutions within the surveillance network are required to report CVD cases using a standardized case report form. In-hospital CVD cases—including those from inpatient, emergency, and outpatient/clinic settings—are recorded by the responsible clinicians or trained staff at each institution. For fatal out-of-hospital CVD events and cross-regional medical incidents, certified physicians from primary healthcare institutions conduct on-site investigations during routine public health activities to complete case reports. Additionally, annual mortality registration data are cross-referenced with the surveillance database, with subsequent verification and supplementary data collection conducted also by primary care physicians. More detailed information can be found in the Supplementary Material.

Major cardiovascular outcomes were monitored, classified using the International Classification of Diseases, 10th Revision (ICD-10) coding system, ensuring precise and standardized categorization. Stroke encompassed a spectrum of conditions, ranging from subarachnoid hemorrhage to ischemic stroke, broadly classified under codes I60, I61, I63,

1 and I64, while explicitly excluding code I62. Coronary heart diseases (CHD) with accurate
2 and reliable diagnoses were selected, including acute myocardial infarction, angina
3 pectoris and sudden cardiac death. Acute myocardial infarction was classified as codes I21-
4 I22, and angina pectoris was represented by code I20, specifically highlighting cases
5 treated with percutaneous transluminal coronary angioplasty (PTCA), stent implantation,
6 and/or coronary artery bypass grafting (CABG). Sudden cardiac death was categorized
7 under code I46.1. Cases with chronic or recovering CVD were explicitly excluded.

8 All districts were defined as urban areas, and all counties, including county-level cities and
9 banners, were defined as rural areas. The classification was based on the participant's
10 habitual residence, defined as having lived in the location for six months or more.

11 To ensure data reliability, the quality of each monitoring site was rigorously evaluated
12 through assessments of validity, completeness, uniqueness, and comparability. Data for
13 cases diagnosed between January 1, 2023, and December 31, 2023, were extracted from
14 the database. After data quality control, a total of 856,357 CVD onset cases were included
15 in this study.

16 *Environmental exposure*

17 We acquired daily meteorological observations in 2023, including maximum, minimum,
18 and mean temperatures, along with relative humidity, from over 2,400 meteorological
19 stations via the China Meteorological Data Sharing Service System (Figure S1). Bilinear
20 interpolation was applied to estimate specific daily environmental exposures for each
21 participant. We also evaluated the results using other interpolation methods, including
22 bicubic and inverse distance weighting. The Pearson correlation coefficients between these
23 results and those obtained using bilinear interpolation were all close to 1, and statistically
24 significant at the 99% confidence level. The extreme heat was defined as the 95th percentile

of daily maximum temperatures. In addition, daily air pollution data for particulate matter less than 2.5 μm in diameter ($\text{PM}_{2.5}$) and maximum 8-h average ozone in 2023 were sourced from the Environmental Monitoring Centers of each county. The surface elevation data was obtained from ETOPO 1 Arc-minute global relief model¹⁵.

Statistical analysis

This study employed a two-stage approach to examine the relationship between high temperature and the incidence of acute CVD. In the first stage, we applied a distributed lag nonlinear model (DLNM)¹⁶ to assess the county-specific effects of daily maximum temperatures on acute CVD incidence. A cross-basis function has been refined to integrate exposure-response and lag-response associations. To address overdispersion, we employed quasi-Poisson regression models, adjusting for confounders such as $\text{PM}_{2.5}$, relative humidity, seasonality, and day-of-week effects. A natural cubic spline with four degrees of freedom (df) was used for both temperature exposure and lag effects.

In the second stage, risk estimates from individual counties were pooled to obtain an overall estimate. A two-level random-effects model was constructed to capture both between- and within-province variability by nesting counties within their respective provinces. We computed best linear unbiased predictions (BLUP)¹⁷ at the county level using the multilevel model. This approach enables geographically clustered counties within the same province to share information effectively with each other.

The BLUP values were used to determine the MIT for each county and at the national level. The MIT corresponds to the temperature associated with the lowest risk of CVD incidence, providing critical insights for identifying vulnerable populations and guiding public health interventions aimed at mitigating heat-related health risks.

1 To examine the delayed effects of high temperature, we analyzed lag periods ranging from
 2 lag 0-2 to lag 0-7 days, and used the lag period with the largest cumulative effect in the
 3 model. Furthermore, we performed stratified analyses to explore the differential effects of
 4 high temperature on various demographic groups. Specifically, we compared the impacts
 5 of extreme heat across genders, age groups, urban versus rural populations, and climate
 6 zones.

7 A series of sensitivity analyses were also performed to assess the robustness of the findings.
 8 First, we tested alternative definitions of extreme heat, using the 90th, 92.5th and 99th
 9 percentiles of daily maximum temperatures. Second, given that previous studies have
 10 shown that ozone¹⁸ and altitude¹⁹ may modify the effect of temperature on CVD, we
 11 additionally introduced ozone and altitude as linear terms in the main model. Furthermore,
 12 we varied the df for meteorological variables from 3 to 5 to check if there was any change
 13 in the results.

14 To quantify the heat-related burden of incident acute CVD, we followed the methodology
 15 that Gasparrini²⁰ developed to calculate AF and AN. AF was computed as

$$16 \quad AF_x = 1 - \exp(-\beta x)$$

17 and AN was calculated as

$$18 \quad AN_x = n \cdot AF_x$$

19 where n is the total number of CVD onset cases and βx represents the exposure-response
 20 coefficient for a given temperature x . Two temperature ranges were considered: 1) all
 21 heat, referring to the temperature effects above the MIT, and 2) extreme heat, referring to
 22 temperatures above the 95th percentile of the daily maximum temperature distribution.
 23 Sub-group analysis was performed for AF, and empirical CIs (95%) were calculated

1 through 1,000 Monte Carlo simulations, assuming a multivariate normal distribution of the
2 BLUP²¹.

3 All statistical analyses in this study were carried out using the “dlnm” and “mixmeta”
4 packages in R Version 4.4.0. Continuous variables were described as means $\pm$ standard
5 deviation (SD), and categorical variables as frequencies and percentages. All tests were
6 two-sided, and a *P*-value of less than 0.05 was considered statistically significant.

7 **Results**

8 *Demographic Characteristics*

9 In this study, we analyzed 856,357 incident acute CVD cases. The majority of cases were
10 male, comprising 58.82% (*n* = 503,780). In terms of age distribution, 68.65% (*n* = 587,973)
11 were elderly (aged ≥ 65 years). The overwhelming majority cases were of Han ethnicity,
12 accounting for 97.60% (*n* = 835,907). Almost half (48.97%, *n* = 410,682) were from rural
13 areas. The cardiovascular outcomes monitored included acute stroke (79.36%, *n* =
14 679,436) and acute CHD (20.64%, *n* = 176,921)(*Table 1*).

15 Given China’s extensive geographical diversity, cases were distributed across various
16 climate zones with distinct characteristics. Nearly half (41.92%, *n* = 358,892) occurred in
17 temperate monsoon climate, characterized by moderate temperatures and seasonal rainfall.
18 Another 54.12% (*n* = 463,267) occurred in subtropical monsoon climate, marked by hot
19 and humid summers. A smaller proportion of cases were located in tropical monsoon
20 climate (0.98%, *n* = 8,399) and in temperate continental climate (3.00%, *n* = 25,688).

21 In China, across all study counties in 2023, the mean daily maximum temperature was
22 22.01°C (ranged from -30.03°C to 41.09°C), with mean relative humidity of 70.64%, PM_{2.5}
23 concentration of 31.30 $\mu\text{g}/\text{m}^3$, and ozone concentration of 97.86 $\mu\text{g}/\text{m}^3$ (*Table 2*).

1 ***Lag pattern of the incidence response at extreme heat***

2 *Figure 1* illustrated the lag effects of extreme heat on CVD incidence, with the
 3 corresponding effect estimates provided in *Table S1*. The cumulative RR of CVD incidence
 4 exhibited an initial upward trend, peaking around lag 0-4 days, and then gradually declined.
 5 At lag 0-6 days, the RR was 1.10 (95% CI 0.95-1.27), indicating that the elevated risk
 6 associated with extreme heat exposure persists for up to one week. These findings
 7 suggested that the impact of elevated temperatures on CVD incidence was acute, with the
 8 highest risk occurring within the initial few days following exposure.

9 ***Temperature-incidence relationship cumulated over 4 days***

10 As shown in *Figure 2*, the pooled relationship between temperature and acute CVD
 11 incidence demonstrated a non-linear association, cumulated over a 4-day lag period. As the
 12 temperature exceeded approximately 30°C, a notable rise in risk was observed. The pooled
 13 RR of CVD incidence associated with extreme heat (95th percentile vs. MIT, around
 14 38.9°C vs. 19.3 °C) was 1.17 (95% CI 1.05-1.30). Subgroup analysis further showed that
 15 the RR of CVD incidence due to extreme heat was higher in individuals aged 65 years and
 16 above (RR = 1.20, 95% CI 1.05-1.37, p<0.01), female (RR = 1.19, 95% CI 1.01-1.40,
 17 p=0.78), the rural population (RR = 1.18, 95% CI 1.01-1.37, p=0.03), and those living in
 18 Temperate Monsoon/Continental climates (RR = 1.25, 95% CI 1.05-1.49, p=0.04),
 19 compared to their respective counterparts (*Figure 2 and Figure 4A*).

20 ***Geographical distribution of the MIT (°C)***

21 The geographical distribution of the MITs varied in China (*Figure 3*), ranging from 16.0 to
 22 26.2 °C. The MIT distribution showed a clear increasing trend from north to south, with all
 23 values exceeding 24°C south of the Tropic of Cancer. This spatial pattern highlighted the

1 variation in temperature-related health risks across different climatic zones in China, with
2 highest MITs concentrated in tropical regions with warmest climates.

3 ***Burden of acute CVD incidence attributable to all heat and extreme heat***

4 In 2023, the AF of acute CVD incidence due to all heat exposures was 3.19% (95% CI
5 1.34%-4.98%), which corresponded to 27,298 CVD onset cases, resulting in a large heat
6 burden. Regarding extreme heat, we found that 0.08% (95% CI 0.02%-0.15%) of the total
7 sample occurred during the 5% hottest days of the year, representing 685 CVD onset cases.
8 *Figure 4* illustrated that rural residents and the elderly were more affected by high
9 temperature and experienced a higher health burden of CVD both during all heat and
10 extreme heat (*Figure 4* and *Table S2*).

11 ***Sensitivity Analyses***

12 Sensitivity analyses were performed to validate our main findings. The association between
13 extreme heat and acute CVD incidence remained consistent when using alternative
14 thresholds to define extreme heat, specifically by replacing the 95th percentile of daily
15 maximum temperature with the 90th, 92.5th and 99th percentiles. Additionally, altering the
16 *df* for meteorological variables from 3 to 5, or extending the main model to include ozone
17 and altitude as covariates to account for potential confounding, did not change the findings
18 substantially, further indicating the robustness of the results(*Table S3*).

19 **Discussion**

20 This study was a national-level investigation in China to investigate the relationship
21 between heat exposure and acute CVD incidence and estimated the heat burden across
22 diverse geographical regions. The study also examined the MIT for CVD incidence,

accounting for both cumulative and delayed health impacts. Our primary findings demonstrate that extreme heat significantly elevates the risk of acute CVD incidence, with the risk peaking within the first four days following exposure. We also found that the heat-related risk and burden varied by age, and urbanization level. These results advanced the existing body of knowledge by emphasizing the broader health impacts of extreme heat exposure on incidence. Importantly, our findings highlighted the urgent need for tailored public health interventions, especially in light of ongoing climate change.

Our findings regarding the association between high temperatures and CVD aligned closely with previous research from different climate zones. Achebak and colleagues¹² conducted a quantitative analysis on the association between heat and CVD hospitalizations in Spain, showing that elevated temperatures slightly increased the risk of hospitalization for diseases of arteries, arterioles, and capillaries (RR: 1.104 [1.038 to 1.175]) and ischemic heart diseases (RR: 1.083 [1.052 to 1.114]). Although the demographic characteristics of patients varied, Achebak's study mainly focused on Whites with a lower proportion of elderly (aged ≥ 65 years), at 18.9%. Similarly, Tao and colleagues²² observed that excess hot days were associated with an increased risk of CVD in China (RR: 1.10 [1.05 to 1.15]). However, Tao's study was limited to data solely from the largest hospitals in 15 cities, excluding rural areas and other provinces.

Previous studies on heat-related health burden have mainly focused on mortality and all-cause incidence, while few studies so far have specifically examined the heat-related health burden of CVD incidence¹⁴. Concentrating on all-cause outcomes could dilute the heat-incidence relationship, making it challenging to detect the specific effects of heat²³. Our study extended this understanding by demonstrating that total and extreme heat contributed to 3.19% and 0.08% of CVD incidence, respectively. Bai and colleagues²⁴ reported that in Ontario, Canada, 1.2% of CHD hospitalizations and 1.8% of stroke hospitalizations were

1 due to heat exposure, with 0.16% and 0.19% attributable to extreme heat. However, their
2 analysis focused solely on hospitalizations (both new and preexisting cases), excluding
3 outpatient settings such as emergency visits. Another study in Shenzhen, a southern
4 Chinese city, found that 2.5% of first-ever stroke incidence was attributable to heat²⁵,
5 underscoring the significant burden of heat on CVD in urban areas. These emphasizes the
6 importance of calculating the AF and AN, which, by translating individual risk
7 (represented by RR) into population-level impact, could be indispensable for public health
8 service evaluation and planning.

9 Another key finding of this study was the geographic variation in MIT, showing a north-
10 south gradient that likely reflects human physiological²⁶⁻²⁸, behavioral and cultural
11 adaptations to the local climate²⁹. Individuals in tropical regions, accustomed to
12 consistently warmer climates, generally exhibit higher MITs, while those in northern areas
13 were more susceptible to the heat's effects. This corresponds with our subgroup analysis,
14 which showed that populations in northern climate zones had higher risk compared to those
15 in more southern climate zones. Although no previous studies have specifically examined
16 the MIT for CVD, similar trends were observed in studies of the minimum mortality
17 temperature. A study found that countries with hotter climates or close proximity to the
18 equator tend to have higher minimum mortality temperatures, suggesting a degree of
19 population adaptation to local climatic conditions³⁰. Yin and colleagues³¹ analyzed 420
20 locations and found that minimum mortality temperatures closely aligned with the most
21 frequently observed local temperatures during the same period. This correlation highlights
22 how long-term climate exposure shapes physiological and behavioral adaptations to
23 environmental conditions. Recognizing these geographic differences highlights the
24 importance of incorporating regional climate characteristics into public health strategies.
25 Our findings confirmed that extreme heat exerts acute and short-term lag effects on
26 cardiovascular health. Consistent with existing literature from China, we observed that the

1 heat effect on CVD was immediate and can persist for up to approximately one week
2 following exposure^{32,33}. This lag effect indicated that individuals might remain at
3 heightened risk for CVD even after extreme heat had dissipated, underscoring the necessity
4 for timely public health interventions during heatwave events and continued monitoring in
5 the days afterward.

6 The findings of our study were in line with previous meta-analyses^{34,35}, which demonstrate
7 that the association between heat and CVD varied by population characteristics and
8 geographical location. Subgroup analyses further revealed disparities in the effects of
9 extreme heat across different demographic and regional groups. Older individuals appeared
10 to be at greater risk, likely due to physiological differences in thermoregulation. Elderly
11 individuals were more vulnerable because of age-related impairments in temperature
12 regulation and the increased prevalence of comorbidities, which amplified their
13 susceptibility to heat-related health impacts^{29,36}. Rural populations also appeared more
14 vulnerable to extreme heat, likely due to limited access to healthcare services and greater
15 exposure to outdoor work. Specifically, many rural residents in China are farmers or
16 laborers and live in underdeveloped housing^{29,37}, which can exacerbate heat exposure. In
17 2023, the average number of air conditioning units in rural areas was 105.7 units per 100
18 households, significantly lower than the 171.7 units per 100 households in urban areas³⁸.

19 Our findings are supported by physiological plausibility. Exposure to extreme heat induces
20 various stress responses in the cardiovascular system. To dissipate heat, the body activates
21 mechanisms such as increased skin blood flow and sweating, which elevate cardiac output
22 to maintain blood pressure and support thermoregulation^{39,40}. Dehydration caused by
23 extreme heat further exacerbates cardiovascular stress by reducing blood volume and
24 decreasing the heart's filling pressure, which can lead to ischemia or infarction⁴¹. Heat
25 exposure can trigger an inflammatory response, with mediators like cytokines disrupting

cellular homeostasis and impairing organ function, thereby exacerbating damage to vulnerable tissues and organs⁴². In addition, extreme heat may also impact cardiovascular health through indirect behavioral and psychological pathways. The E(e)SEEDi lifestyle framework integrates environmental exposures with four core domains of health—Sleep, Emotion, Exercise, and Diet⁴³. According to this framework, environmental stressors such as extreme heat can compromise the quality of these essential lifestyle factors, thereby increasing cardiovascular vulnerability.

Although this study focused on the adverse effects of extreme ambient heat exposure, it is important to acknowledge that not all forms of heat exposure are harmful. For instance, controlled and time-limited exposures such as Finnish sauna bathing have been associated with cardiovascular benefits, including improved vascular function, reduced blood pressure, and lower risk of cardiovascular events and mortality⁴⁴.

This study has several notable advantages. First, it stands as one of the largest studies to date to examine the effects of extreme heat on CVD incidence in China, covering multiple counties across a range of climate zones. Such extensive cases, combined with broad geographic coverage, allow for a more comprehensive and robust assessment of the relationship between high temperature and CVD risk. Secondly, this study explored MIT variations across different climate zones, providing novel insights into CVD prevention and serving as a valuable reference for region-specific public health interventions. Additionally, few studies have conducted subgroup analyses that differentiate urban and rural populations^{6,12,45}, and this study addressed this gap. By identifying specific vulnerable groups, such as rural residents and elderly individuals, this study contributed to the design of targeted adaptation strategies for more effective health outcomes.

However, this study has several limitations. First, as an ecological study, this study was subject to ecological fallacies, and caution was needed when generalizing the findings to

1 other contexts or regions. Second, although we controlled for environmental confounding
2 factors such as humidity and air pollution, individual-level factors such as lifestyle and
3 behavior still influence the results. Additionally, since most individuals spent the majority
4 of their time indoors^{46,47}, relying on outdoor temperature observations might not fully
5 reflect actual heat exposure at the individual level. Beyond coronary heart disease and
6 stroke, extreme heat may also increase the risk of other major adverse cardiovascular
7 events (MACE), including arrhythmias and heart failure. However, as these conditions are
8 not included in the current scope of the CVD surveillance system in China, they were not
9 assessed in this analysis. Furthermore, the study was based on data from one single year,
10 limiting its ability to assess long-term trends over time. Future research should address
11 these limitations by incorporating more precise individual-level exposure data, exploring
12 compound exposure to other environmental risk factors, using composite
13 biometeorological indicators such as net effective temperature⁴⁸, and examining long-term
14 trends across diverse global contexts to provide a more comprehensive understanding of
15 heat-related cardiovascular risks.

16 In conclusion, our study demonstrated that extreme heat significantly elevated the risk of
17 acute CVD incidence in China, with notable variations across geographic regions and
18 demographic groups. These findings underscore the urgent need of implementing targeted
19 measures to mitigate heat exposure, particularly for vulnerable populations. Such as,
20 establishing early warning systems and enhancing public awareness can enable
21 communities to better anticipate and respond to extreme heat events. And, developing
22 tailored interventions to reduce heat exposure for high-risk populations is essential.
23 Together, these strategies provide a comprehensive approach to reduce the adverse health
24 effects of extreme heat on the Chinese population, mitigate CVD risk, and enhance
25 community resilience in the context of ongoing climate change.

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8 Conflicts of interest

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

12 Authors' Contributions

13 ZW had full access to all of the data in the study and takes responsibility for the integrity
14 of the data and the accuracy of the data analysis. ZW: conceptualization, data curation,
15 methodology, review & editing, supervision, funding acquisition; XP: methodology, formal
16 analysis, writing-original draft preparation; HT: methodology and writing-original draft
17 preparation; CZ, YZ, MS, QJ, SL, XW, YT, XC, YT, XY, NT, CC, YZ, FL and JY:
18 investigation.

20 Data Availability Statement

21 The health data underlying this article will be shared on reasonable request to the
22 corresponding author. The environmental data used in this study are publicly available from

their respective sources. The daily meteorological observations from over 2,400 meteorological stations were obtained from the China Meteorological Data Sharing Service System (CMDC, available at <http://data.cma.cn>). Air pollution data were sourced from the Environmental Monitoring Centers of each county in China (available at <https://quotsoft.net/air/>). Surface elevation data were derived from the ETOPO 1 Arc-Minute Global Relief Model, provided by the National Centers for Environmental Information (NOAANCEI, available at <https://www.ncei.noaa.gov/products/etopo-global-relief-model>).

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1 **Figure legends**

2 **Figure 1.** Cumulative Lag Effects of Extreme Heat on CVD Incidence Over One Week. RR, relative
3 risk. The effect of extreme heat was summarized as the relative risk (RR) of incidence at the 95th
4 temperature percentiles vs. the minimum incidence temperature (MIT).

5 **Figure 2.** Temperature-Incidence Relationship and RR of CVD Over a 4-Day Lag Period. (A) Pooled
6 Non-Linear Relationship Between Temperature and CVD Incidence. (B) RR of CVD Incidence by
7 Gender. (C) RR of CVD Incidence by Age Group. (D) RR of CVD Incidence by Urbanicity. (E) RR of
8 CVD Incidence by Climate Zone. RR, relative risk. Dashed vertical lines in each panel indicate the
9 threshold for extreme heat, defined as the 95th percentile of daily temperature distribution. Asterisk (*)
10 indicates $P < 0.05$, and double asterisk (**) indicates $P < 0.01$ for differences in RR between extreme
11 heat and CVD incidence.

12 **Figure 3.** Geographical Distribution of Minimum Incidence Temperature Across China.

13 **Figure 4.** Subgroup Analysis by Age, Sex, Urbanicity and Climate Zone (A) RR and 95%CI of CVD
14 Incidence Associated with Extreme Heat. (B) AF and 95%CI of Extreme Heat Exposures. (C) AF and
15 95%CI of All Heat Exposures. RR, relative risk. AF, Attributable fraction.

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1 **Tables**2 **Table 1. Demographic Characteristics and CVD Outcomes of the Study Population**

Characteristics	Female n=352,577 (%)	Male n=503,780 (%)	Total n=856,357(%)	P-Value	
Age					
18-64	81,159 (23.02)	187,225 (37.16)	268,384 (31.35)	<0.001	
≥65	271,418 (76.98)	316,555 (62.84)	587,973 (68.65)		
Ethnic					
Han	344,968 (97.84)	490,939 (97.45)	835,907 (97.60)	<0.001	
National minority	7,609 (2.16)	12,841 (2.55)	20,450 (2.40)		
Urbanity					
Urban	177,252 (50.27)	268,423 (53.28)	445,675 (52.03)	<0.001	
Rural	175,325 (49.73)	235,357 (46.72)	410,682 (48.97)		
Career					
White collar	12,099 (3.43)	24,180 (4.80)	36,279 (4.23)	<0.001	
Administrator	3,702 (1.05)	13,075 (2.60)	16,777 (1.96)		
Blue collar	3,864 (1.10)	5,526 (1.10)	9,390 (1.10)		
Other	221,880 (62.93)	306,746 (60.89)	528,626 (61.76)		
Missing	111,032 (31.49)	154,253 (30.62)	265,285 (30.96)		
Marital status					
Married/remarried /living together	214,743 (60.91)	324,193 (64.35)	538,936 (62.98)		<0.001
Separated divorced	1,481 (0.42)	3,790 (0.75)	5,271 (0.62)		
Single	3,735 (1.06)	12,556 (2.49)	16,291 (1.90)		
Widowed	23,844 (6.76)	12,192 (2.42)	36,036 (4.21)		
Missing	108,774 (30.85)	151,049 (29.98)	259,823 (30.32)		
Diseases					
CHD	61,447 (17.43)	115,474 (22.92)	176,921 (20.64)	<0.001	
Stroke	291,130 (82.57)	388,306 (77.08)	679,436 (79.36)		

Climate zones				
Temperate Monsoon Climate	150,835 (42.84)	208,057 (41.30)	358,892 (41.92)	
Subtropical Monsoon Climate	187,545 (53.33)	275,722 (54.62)	463,267 (54.12)	
Tropical Monsoon Climate	3,418 (0.97)	4,981 (0.99)	8,399 (0.98)	
Temperate Continental Climate	10,086 (2.86)	15,602 (3.10)	25,688 (3.00)	<0.001

1 * *P*-values represent comparisons between groups for each characteristic.

2 **Table 2. Distributions of Daily Ambient Environment Conditions in China in 2023**

Environmental factors	Mean	Minimum	Q1	Q2	Q3	Maximum
Maximum temperature(°C)	21.00	-30.03	14.20	22.91	29.19	41.09
Relative humidity(%)	70.64	5.80	56.29	75.72	87.99	100.00
PM _{2.5} (μ g/m ³)	31.30	1.00	14.50	24.50	39.80	334.90
Ozone(μg/m ³)	97.86	5.90	71.80	91.70	120.10	274.00
altitude(m)	232.43	1.70	18.30	49.70	244.10	2524.80

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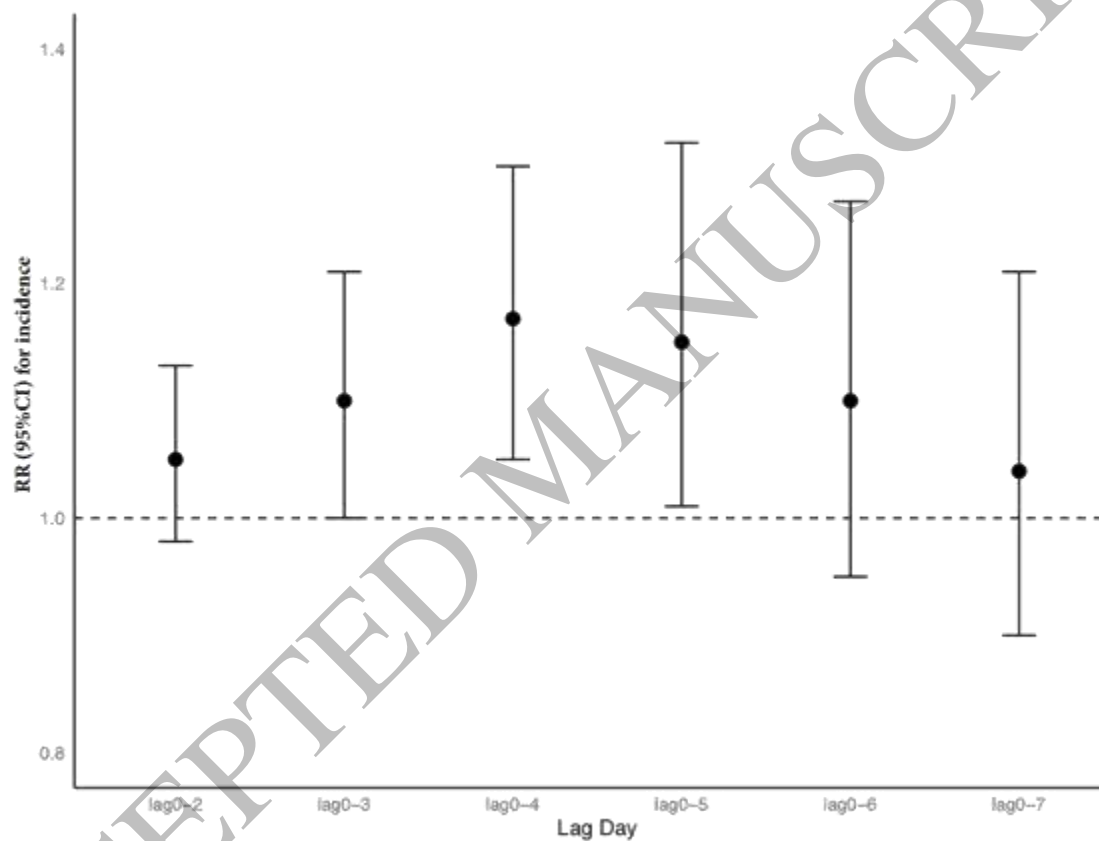
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2 **Figures**

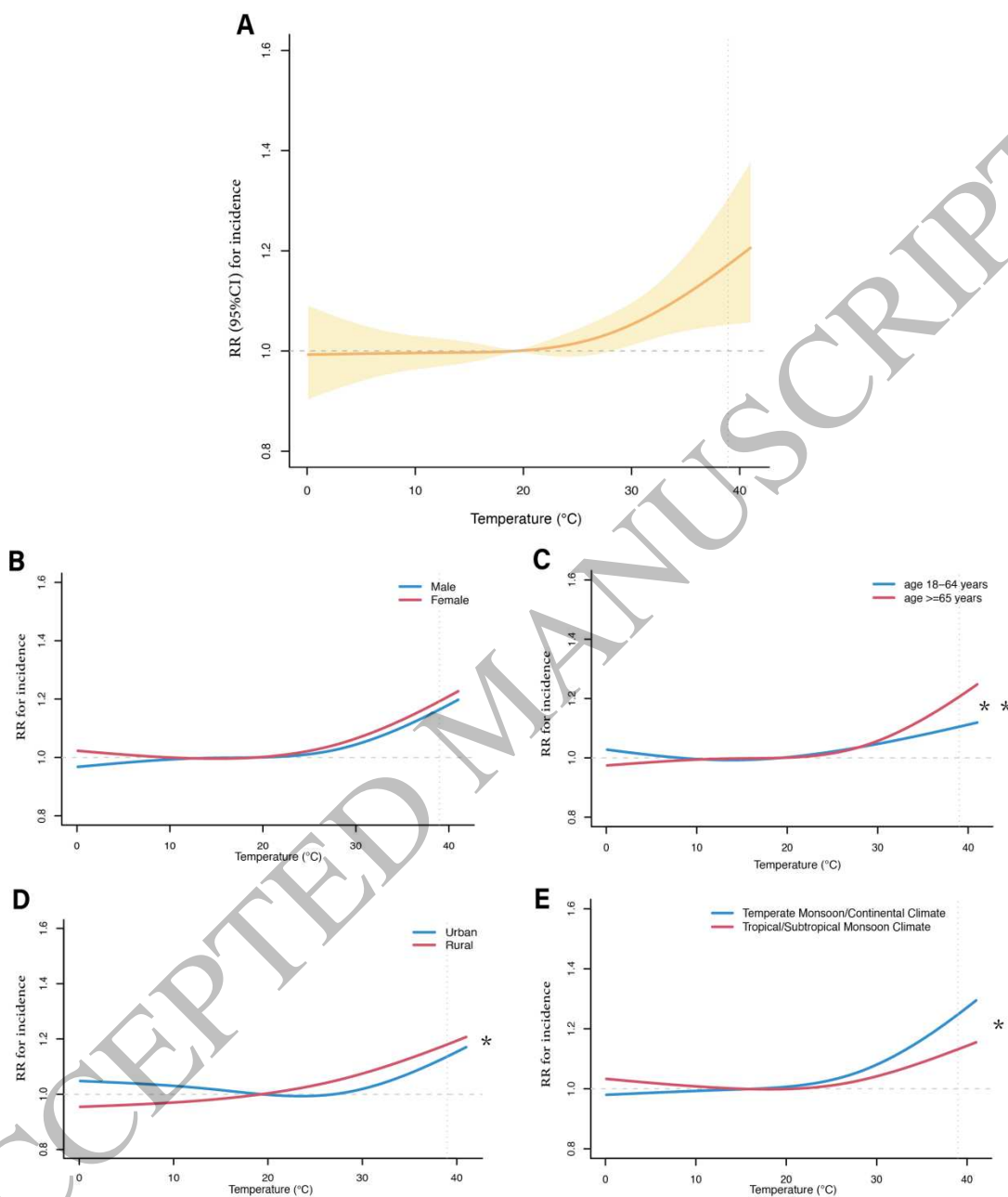
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5 **Figure 1 Cumulative Lag Effects of Extreme Heat on CVD Incidence Over One Week.** RR, relative
 6 risk. The effect of extreme heat was summarized as the relative risk (RR) of incidence at the 95th
 7 temperature percentiles vs. the minimum incidence temperature (MIT).

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2 **Figure 2 Temperature-Incidence Relationship and RR of CVD Over a 4-Day Lag Period. (A)**
3 **Pooled Non-Linear Relationship Between Temperature and CVD Incidence. (B) RR of CVD**
4 **Incidence by Gender. (C) RR of CVD Incidence by Age Group. (D) RR of CVD Incidence by**
5 **Urbanicity. (E) RR of CVD Incidence by Climate Zone. Dashed vertical lines in each panel indicate**
6 **the threshold for extreme heat, defined as the 95th percentile of daily temperature distribution. Asterisk**
7 **(*) indicates $P < 0.05$, and double asterisk (**) indicates $P < 0.01$ for differences in RR between extreme**
8 **heat and CVD incidence.**

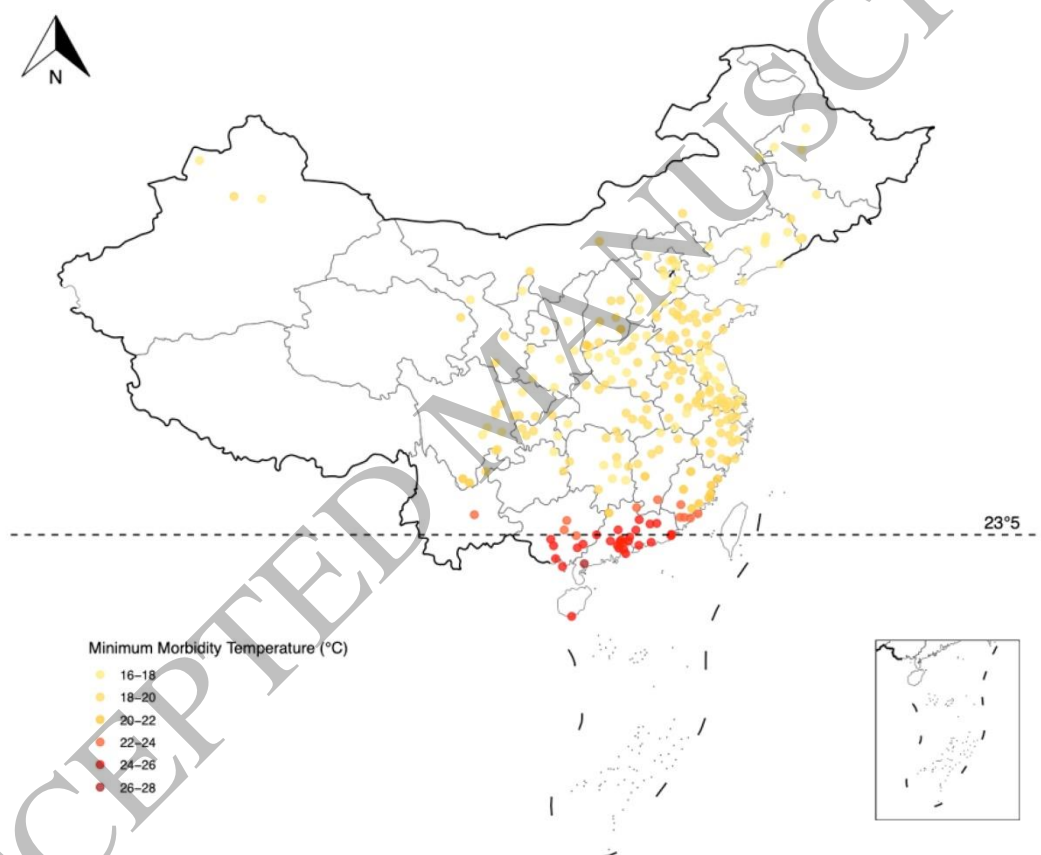


Figure 3 Geographical Distribution of Minimum Incidence Temperature Across China

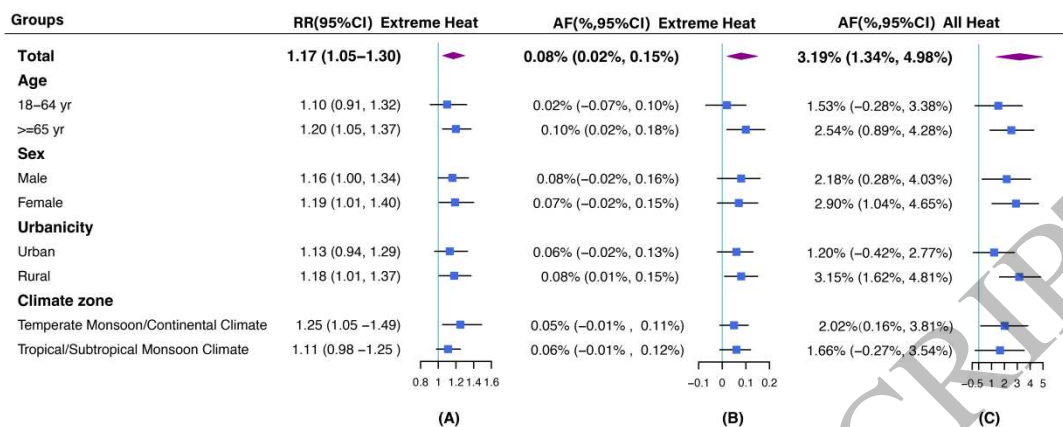
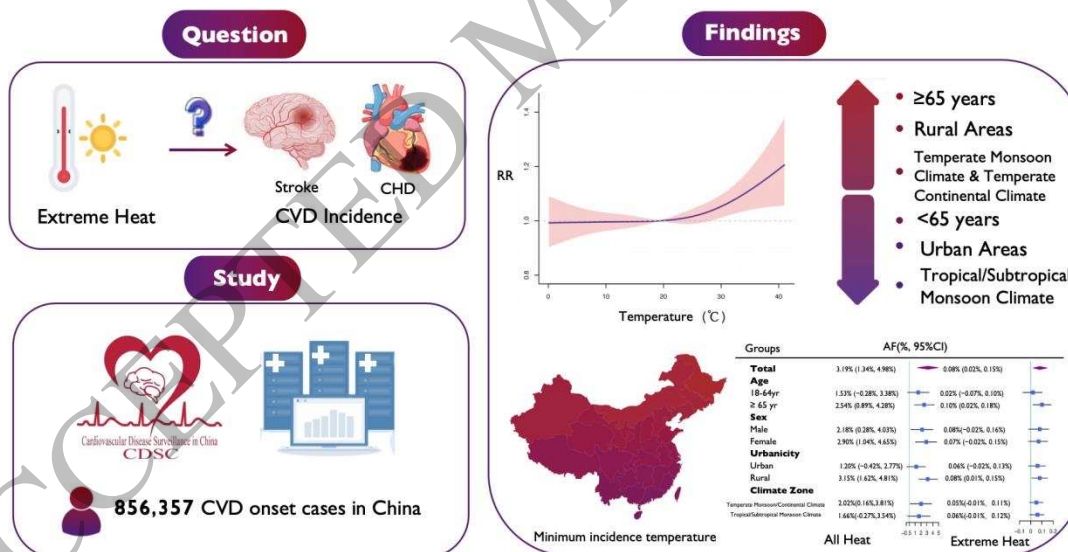


Figure 4 Subgroup Analysis by Age, Sex, Urbanicity and Climate Zone (A) RR and 95%CI of CVD Incidence Associated with Extreme Heat. (B) AF and 95%CI of Extreme Heat Exposures. (C) AF and 95%CI of All Heat Exposures. RR, relative risk. AF, Attributable fraction.



Graphical abstract