



9 Abstract

10 Heat stress is intensifying across Southeast Asia under global warming, yet the relative
11 influences of atmospheric warming and moistening across timescales remain insufficiently
12 quantified. This study investigates the thermodynamic drivers of warm-season (April–October) heat-
13 stress intensity, measured by the daily maximum wet-bulb globe temperature (*WBGT_{max}*), and
14 frequency, defined as the annual number of extreme heat-stress days (*N_{xday}*), across Southeast Asia
15 and its 20 climatic sub-regions. Using observations together with dynamically downscaled
16 CORDEX–SEA simulations, we apply a unified attribution framework to separate the effects of air
17 temperature, specific humidity, and residual nonlinear processes on historical (1985–2014) trends,
18 interannual variability, and projected late-century changes (2071–2100 relative to 1985–2014) under
19 SSP5–8.5. Historical increases in *WBGT_{max}* and *N_{xday}* are dominated by temperature across much
20 of the region. However, humidity already provides important amplification in monsoon-influenced
21 lowlands, including Indochina, the Philippines, and parts of the Malay Peninsula. Future projections
22 indicate a coherent basin-wide shift toward compound warm–humid conditions. In many monsoon
23 regions, rising moisture contributes roughly 30–50 % of the increase, consistent with enhanced
24 atmospheric water-holding capacity in a warmer climate. In contrast, interannual variability,
25 particularly over the Maritime Continent, is strongly governed by nonlinear temperature–humidity
26 interactions, which generate substantial unexplained components. Because *N_{xday}* depends on
27 threshold exceedance, it shows a stronger amplification of future change than *WBGT_{max}*, even
28 though its long-term evolution remains primarily temperature controlled. Overall, the results
29 demonstrate that heat-stress escalation in Southeast Asia increasingly reflects rising atmospheric
30 moist enthalpy rather than dry-bulb warming alone, underscoring the need for adaptation strategies
31 that explicitly consider both temperature and humidity.

32 **Keywords:** *heat stress; wet-bulb globe temperature (WBGT); global warming; temperature–*
33 *humidity interactions; thermodynamic decomposition; Southeast Asia.*



34 **1. Introduction**

35 Southeast Asia (SEA), situated in the tropical Asia–Pacific region, includes both mainland and
36 maritime territories of eleven countries and is home to more than 700 million people. The climate is
37 characterized by persistently high temperature and humidity, with rainfall strongly regulated by the
38 Asian monsoon system (Räsänen et al., 2016). Despite its broadly tropical setting, SEA exhibits
39 pronounced spatial contrasts and strong seasonality in both precipitation and temperature due to
40 interactions among large-scale circulation, land–sea thermal contrasts, and complex topography
41 (Chang et al., 2005; Juneng et al., 2016). To better represent this diversity, SEA has been partitioned
42 into 20 climatic sub-regions within the SEA regional climate downscaling/coordinated regional
43 climate downscaling experiment–SEA (SEACLID/CORDEX–SEA) project (Juneng et al., 2016;
44 Cruz et al., 2017; Tangang et al., 2020).

45 SEA is widely regarded as one of the regions most exposed to climate change impacts (Collins
46 et al., 2013), while mitigation and adaptation capacities vary substantially among countries
47 (Overland et al., 2021). Among climate-related hazards, increasing heat stress presents serious
48 threats to public health, outdoor labor, and economic productivity (Barriopedro et al., 2011; Dunne et
49 al., 2013; Kjellstrom et al. 2016, 2018). In contrast to many mid-latitude regions, where extreme heat
50 stress is often dominated by dry-bulb temperature, heat stress in SEA is strongly governed by the
51 combined influence of air temperature and atmospheric moisture (e.g., Zhou et al., 2024; Kushwaha
52 et al., 2026). The wet-bulb globe temperature (WBGT) is therefore widely used to quantify human
53 heat exposure, particularly in climate-change studies, as it explicitly captures this thermodynamic
54 coupling between temperature, humidity, radiation, and wind (Yaglou and Minard, 1957, De Freitas
55 and Grigorieva, 2007).

56 Consistent increases in WBGT have been reported under global warming at both global and
57 regional scales (e.g., Im et al., 2017, Lee and Min, 2018; Schwingshackl et al., 2021, Mohammad
58 and Weng, 2025, Kushwaha et al., 2026). However, the relative importance of rising temperature



59 versus increasing atmospheric moisture remains insufficiently constrained. Quantifying these
60 contributions is essential for adaptation planning, since temperature-dominated heat stress favors
61 strategies such as shading or insulation, whereas humidity-driven stress highlights the need for
62 ventilation and the limits of evaporative cooling. Furthermore, the mechanisms governing long-term
63 changes may differ from those responsible for interannual variability. While anthropogenic forcing
64 produces a persistent warming trend, climate mode such as the El Niño–Southern Oscillation
65 (ENSO) can strongly modulate regional moisture conditions, generating short-lived yet intense heat
66 stress anomalies (Raymond et al., 2020). This issue is particularly relevant for SEA, where ENSO
67 teleconnections exert a major control on regional hydroclimate (Nguyen-Le, 2024).

68 SEA thus provides a valuable setting for investigating the thermodynamic controls of heat stress.
69 The maritime environment maintains high background humidity, meaning that relatively small
70 temperature increases can drive WBGT toward thresholds critical for human tolerance (Buzan and
71 Huber, 2020). Although many studies have assessed projected changes in heat stress magnitude (e.g.
72 Manimaran et al., 2025; Nguyen-Le et al., 2026), fewer have systematically separated the effects of
73 temperature and moisture or compared the drivers of mean-state change with those of interannual
74 variability across spatial scales.

75 This study addresses that gap by quantifying the respective roles of near-surface air temperature
76 (T) and specific humidity (q) in shaping heat stress across SEA. We analyze the historical period
77 (1985–2014) and late-century projections (2071–2100) under the SSP5–8.5 scenario (Eyring et al.,
78 2016; O’Neill et al., 2017 using observations together with dynamically downscaled CORDEX–SEA
79 simulations. A unified attribution framework is applied to evaluate contributions of T and q to (1)
80 historical trends, (2) interannual variability, and (3) climatological mean-state changes between
81 historical and future climates. The results provide spatially explicit insight into where and why
82 WBGT changes occur, offering information relevant for risk assessment and adaptation planning
83 across SEA and its 20 climatic sub-regions.



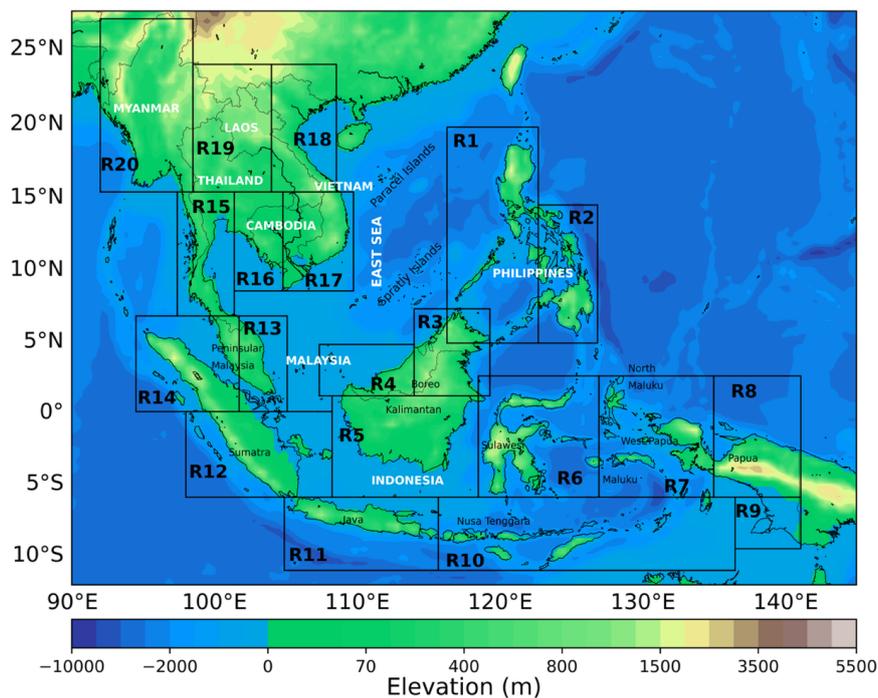
84 The remainder of the paper is organized as follows. Section 2 describes the study region,
85 datasets, WBGT estimation, and attribution methods. Section 3 presents the results, and Section 4
86 concludes with discussion and implications.

87 **2. Study region, datasets, and methodology**

88 **2.1 Study region**

89 Figure 1 shows the SEA domain (12° S–27.5° N, 90° E–145° E) and the 20 climatic sub-regions
90 defined by Juneng et al. (2016). Although much of SEA lies within the tropical belt, where annual
91 mean temperature typically exceeds 25 °C, substantial sub-regional contrasts in circulation and
92 rainfall lead to marked differences in humidity and, consequently, heat-stress conditions.

93 In the northern part of SEA, including northern Indochina (R18–R19), Myanmar (R20), and the
94 Philippines (R01–R02), boreal-winter and early spring northeasterly cold surges from East Asia
95 bring episodic cooling and drying (Chang et al., 2005), temporarily reducing heat stress. During the
96 remainder of the year, precipitation becomes the primary control on moisture availability and thus on
97 the seasonal evolution of heat stress. Mainland SEA (R15–R20) and the Philippines are strongly
98 influenced by the Asian summer monsoon (Matsumoto 1997; Wang and LinHo 2002; Nguyen-Le
99 2023). During boreal summer, enhanced humidity substantially intensifies heat stress even where
100 temperature variations are modest. Over much of the Maritime Continent (R04–R12), peak rainfall
101 instead occurs during the boreal winter monsoon (Juneng et al. 2016), sustaining high humidity and
102 favoring near year-round heat exposure. Other areas, including Borneo (R03), western Malaysia
103 (R13), and northern Sumatra (R14), experience rainfall maxima during boreal autumn (Nguyet-Le
104 2023). In these locations, the concurrence of elevated humidity and persistently high temperature
105 produces pronounced seasonal peaks in moist heat stress.



106

107 **Figure 1:** Southeast Asia (SEA) domain and its 20 climatic sub-regions (outlined by black boxes) following
 108 the SEACLID/CORDEX–SEA framework. Shading indicates topography from the Global 30 Arc-Second
 109 Elevation (GTOPO30) dataset (EROS, 2017).

110 **2.2 Heat-stress indices and thresholds**

111 Heat stress is evaluated using the wet-bulb globe temperature (WBGT), a composite index that
 112 integrates the effects of air temperature, atmospheric moisture, wind speed, and radiative heat load.
 113 WBGT has been widely applied in occupational health and climate-impact assessments because it
 114 provides a direct measure of environmental constraints on human thermoregulation (De Freitas and
 115 Grigorieva, 2007).

116 WBGT is formulated as a weighted combination of the natural wet-bulb temperature (T_w), black
 117 globe temperature (T_g), and dry bulb temperature (T_a). T_w represents evaporative cooling under
 118 ambient conditions and depends mainly on temperature and humidity. T_g reflects radiative heat gain
 119 from shortwave and longwave fluxes and is additionally influenced by wind speed. T_a corresponds to



120 the shaded air temperature. For outdoor daytime environments with solar exposure, WBGT is
 121 defined as

$$122 \quad WBGT_{outdoor} = 0.7T_w + 0.2T_g + 0.1T_a \quad (1)$$

123 where the weighting emphasizes the dominant role of evaporative cooling limitations in
 124 determining heat strain.

125 Because boreal-winter cold surges substantially reduce heat stress across large parts of Southeast
 126 Asia, the analysis concentrates on the warm season from April to October. WBGT is calculated
 127 following the physically based heat-balance model of (2008), implemented using the open-source
 128 Python framework described by Kong and Huber (2022). Heat-stress intensity is represented by
 129 $WBGT_{max}$, defined as the daily maximum value of outdoor WBGT (Equation 1).

130 **Table 1.** Summary of wet-bulb globe temperature (WBGT) Thresholds (Based on ISO 7243:2017).

Risk Level	WBGT Threshold	Work/Rest Allocation	ISO 7243:2017 Context
Level 1	28 °C	100% Work	Reference Limit (Moderate Work): Safe for sustained 8-hour moderate labor.
Level 2	30 °C	50% Work / 50% Rest	Action Limit: Maximum limit for light work; requires 30 min rest/hour for moderate work.
Level 3	32 °C	25% Work / 75% Rest	High Strain: Severe risk for moderate labor; requires dominant rest periods.
Level 4	>33 °C	No sustained work	Ceiling: Threshold where environmental heat exceeds human cooling capacity for labor.

131

132 Changes in heat-stress indices alone do not fully capture societal impacts, as their relevance
 133 depends on physiologically meaningful thresholds. To translate climatic changes into implications
 134 for human health and labor productivity, we adopt guidance from ISO 7243:2017, which links



135 WBGT levels to physiological strain and work–rest requirements (ISO 2017). An extreme heat-stress
136 day is defined as a day on which $WBGT_{max}$ exceeds 30 °C. This threshold corresponds to the upper
137 reference limit for light manual work and a level of high physiological strain for moderate work, at
138 which allowable work time typically must be reduced by about 50% per hour to maintain core body
139 temperature below 38 °C. Heat-stress frequency is quantified using N_{xday} , defined as the annual
140 number of extreme heat-stress days.

141 Atmospheric moisture is represented by specific humidity (q) rather than relative humidity,
142 which is commonly used in heat-stress analyses (e.g. Zhou *et al.* 2024). Specific humidity q is
143 preferred because it is a conserved thermodynamic variable and allows a clearer physical separation
144 between temperature-driven and moisture-driven influences on WBGT. In contrast, relative humidity
145 is strongly dependent on T and often shows limited change under global warming (Douville *et al.*,
146 2022), which can obscure attribution of the underlying drivers.

147 **2.3 Datasets and preprocessing**

148 Historical heat stress is evaluated using the ECMWF (European Centre for Medium-Range
149 Weather Forecasts) ERA5 reanalysis (Hersbach *et al.* 2020) for 1985–2014. Future conditions are
150 derived from regional climate projections produced within CORDEX–SEA (Tangang *et al.*, 2020;
151 Ngo-Duc *et al.*, 2024), which dynamically downscale simulations from Coupled Model
152 Intercomparison Project (CMIP6; Eyring *et al.*, 2016) to a horizontal resolution of 25 km.

153 Two driving global climate models, NorESM2-MM and CNRM-ESM2-1, are dynamically
154 downscaled using the non-hydrostatic RegCM4-NH regional climate model (Coppola *et al.*, 2021).
155 Model T , q , wind speed, surface pressure, and radiative fluxes are archived at 3-hourly interval. . For
156 consistency with the observational reference, all variables are interpolated onto the 0.25° ERA5 grid
157 prior to WBGT calculation.



158 Although projections are available for multiple Shared Socioeconomic Pathways (SSP1–2.6,
159 SSP2–4.5, and SSP5–8.5) covering 1985–2100, the present analysis emphasizes late-century change
160 (2071–2100) under SSP5–8.5 in order to characterize upper-bound heat-stress risk.

161 Systematic model biases are adjusted using the multivariate bias-correction method based on N-
162 dimensional probability density functions (MBCn). The correction is applied individually to T , q ,
163 wind speed, surface pressure, and shortwave radiation for each regional simulation. By preserving
164 inter-variable dependence, this approach improves consistency in compound indices such as WBGT,
165 which depend jointly on thermodynamic and radiative conditions (Qiu et al. 2023; Zhou et al. 2024).
166 After bias correction, the ensemble mean of the two regional simulations is used to derive WBGT
167 and associated heat-stress metrics.

168 **2.4 Attribution methodology**

169 To attribute the relative roles of T and q in driving WBGT changes, we employ a unified
170 attribution framework composed of three complementary approaches. Each method targets a
171 different aspect of variability: (1) long-term trends, (2) interannual fluctuations, and (3) differences
172 in climatological means between historical and future climates. Although wind speed and radiation
173 influence WBGT through the physical heat-balance model, they are not attributed explicitly here.
174 This design isolates the dominant thermodynamic controls and enables a clearer interpretation of
175 temperature versus moisture contributions.

176 (1) *Trend decomposition*: Long-term trends in $WBGT_{max}$ (or N_{xday}) are attributed using a first-
177 order linear decomposition based on the chain rule of calculus. This approach assumes that temporal
178 changes in $WBGT_{max}$ can be approximated as the linear sum of trends in its driving variables,
179 weighted by their local sensitivities. At each grid point, the linear trend in $WBGT_{max}$ is expressed as

$$180 \quad \frac{d(WBGT_{max})}{dt} \approx \beta_T \frac{dT}{dt} + \beta_q \frac{dq}{dt} + \varepsilon \quad (2)$$



181 where $\frac{dX}{dt}$ denotes the linear trend (slope) of variable X , estimated using ordinary least squares
182 (OLS) regression. The coefficients β_T and β_q represent the partial sensitivities of $WBGTmax$ to T and
183 q , respectively, obtained from a multiple linear regression of $WBGTmax$ onto T and q . The term
184 $\beta_T \frac{dT}{dt}$ and $\beta_q \frac{dq}{dt}$ quantify the contributions of temperature warming and atmospheric moistening (or
185 drying), while ε represents the residual component not explained by linear T - q effects.

186 (2) *Variability attribution*: To quantify the relative importance of T and q in controlling
187 interannual variability in $WBGTmax$ (or $Nxday$), we employ Shapley value regression based on
188 dominance analysis. This method evaluates the marginal contribution of each predictor to the model
189 coefficient of determination (R^2) across all possible predictor combinations. The relative importance
190 of a predictor x_k is defined as

$$191 \quad I(x_k) = \text{Average}(R_{model\ with\ x_k}^2 - R_{model\ without\ x_k}^2) \quad (3)$$

192 where the averaging is performed over all possible sub-model permutations. Based on these
193 contributions, the total interannual variance of $WBGTmax$ is approximately decomposed as

$$194 \quad \sigma_{WBGTmax} \approx \sigma_T + \sigma_q + \sigma_{Unexplained} \quad (4)$$

195 where each term represents the fraction of interannual variance attributable to T variability, q
196 variability, and unexplained processes, respectively. This approach explicitly quantifies the relative
197 influence of thermodynamic drivers on year-to-year $WBGTmax$ fluctuations, including those
198 associated with large-scale climate variability such as ENSO.

199 (3) *Mean-state change attribution*: To attribute changes in the mean climatological state of
200 $WBGTmax$ (or $Nxday$) between the historical (1985–2014) and future (2071–2100) periods, we apply
201 the Oaxaca–Blinder decomposition. This method partitions the change in mean WBGT into
202 components explained by changes in the mean values of the driving variables and an unexplained
203 (residual) term

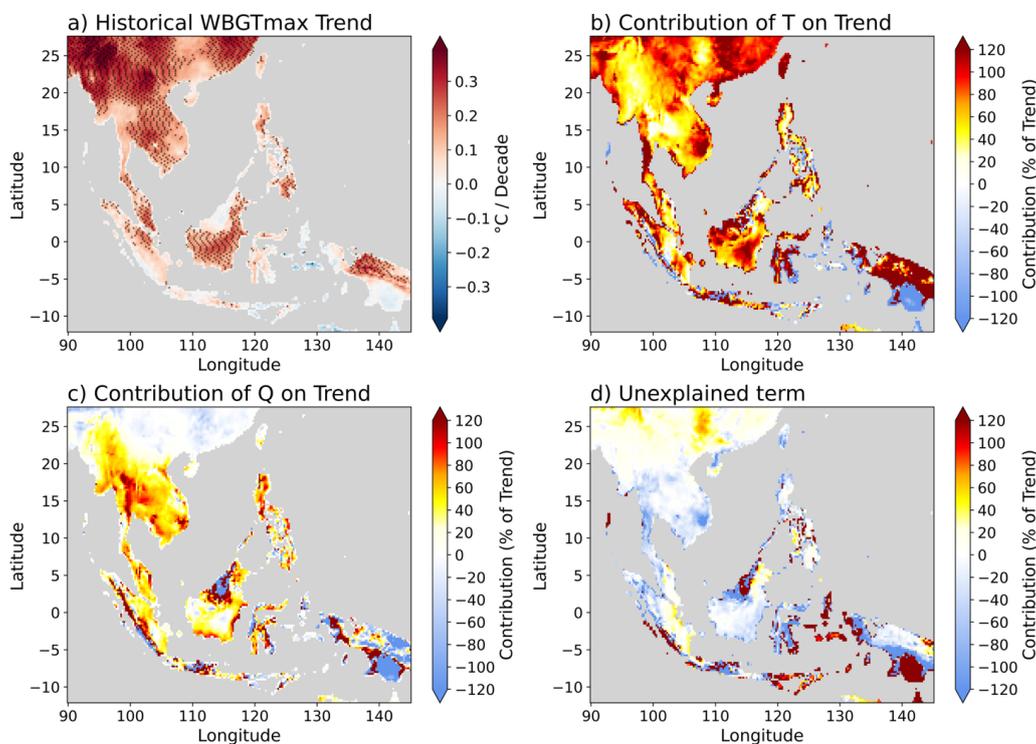


204
$$\overline{\Delta WBGTmax} = \overline{WBGTmax_{fut}} - \overline{WBGTmax_{hist}} \approx \beta_T(\bar{T}_{fut} - \bar{T}_{hist}) + \beta_q(\bar{q}_{fut} - \bar{q}_{hist}) + \varepsilon \quad (5)$$

205 where \bar{X}_{hist} and \bar{X}_{fut} denote the temporal means of variable X during the historical and future
 206 periods, respectively. The coefficients β_T and β_q are pooled regression coefficients derived from the
 207 combined historical and future datasets. The residual term ε captures contributions not explained by
 208 mean changes in T and q , including nonlinear responses and modifications of the WBGT–driver
 209 relationship.

210 **3. Results**

211 **3.1 Drivers of historical (1985–2014) heat stress intensity**



212

213 **Figure 2:** Spatial distribution of historical long-term trends in warm-season (April–October) daily maximum
 214 wet-bulb globe temperature (WBGTmax) over Southeast Asia during 1985–2014. (a) Linear trend in
 215 WBGTmax (°C decade⁻¹); stippling denotes statistical significance at $p < 0.05$. (b–d) Relative contributions



216 (%) of near-surface air temperature (T), specific humidity (q), and the unexplained (residual) component to
217 the $WBGT_{max}$ trend.

218 Figure 2a displays the spatial distribution of linear trends in warm-season (April–October) daily
219 maximum $WBGT$ ($WBGT_{max}$) during the historical period (1985–2014). Positive trends occur
220 across nearly all of SEA, indicating a widespread intensification of heat-stress conditions.

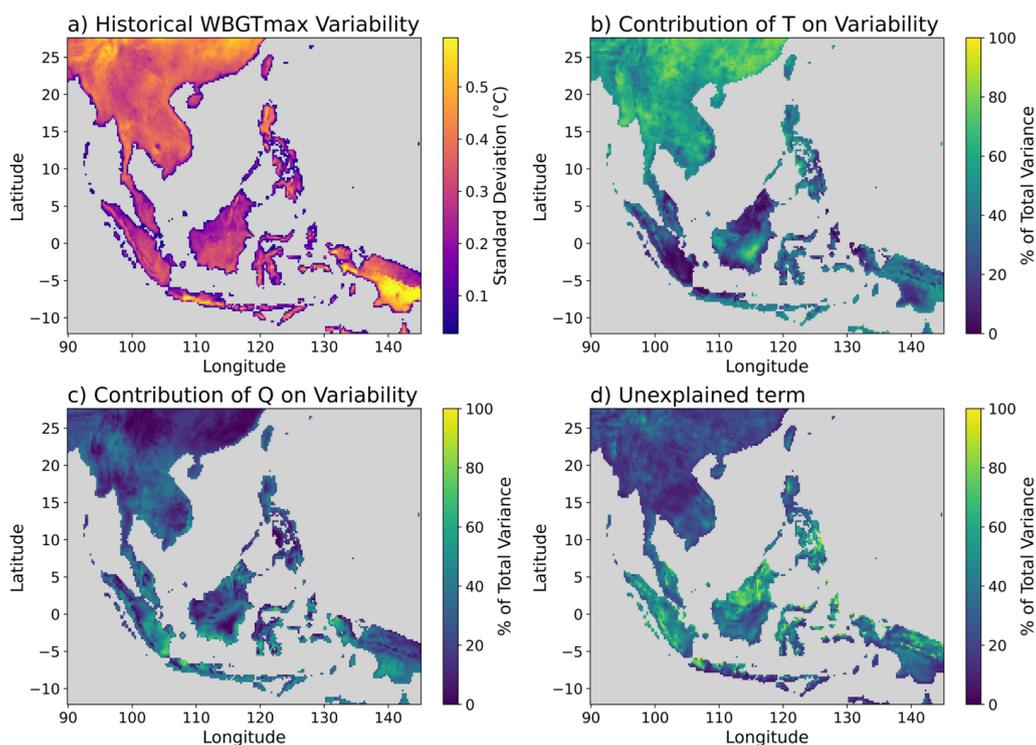
221 Statistically significant increases of around $0.2\text{ }^{\circ}\text{C decade}^{-1}$, and locally higher, are concentrated over
222 mainland Southeast Asia, the Malay Peninsula, parts of the eastern Philippines, and regions of the
223 Maritime Continent such as Kalimantan and Papua. In contrast, weaker and more fragmented trends
224 appear over some equatorial areas. Overall, the magnitude of the trends tends to increase toward
225 higher latitudes and over elevated terrain, suggesting modulation by both geographic setting and
226 topography.

227 The driver decomposition reveals pronounced spatial contrasts in the relative contributions of air
228 temperature (T) and specific humidity (q) to historical $WBGT_{max}$ trends (Figs. 2b–d). These patterns
229 agree with the sub-regional synthesis (Fig. 4a) and with the corresponding trends in T and q (Fig.
230 S1). A clear distinction emerges between monsoon-influenced regions of northern SEA (including
231 lowland Indochina, the Philippines, and parts of Malay Peninsula) and the equatorial Maritime
232 Continent. Over much of the equatorial Maritime Continent, $WBGT_{max}$ trends are primarily
233 controlled by T (Fig. 2b), indicating a predominantly temperature-controlled response to long-term
234 warming. Elevated and mountainous regions, including northern Indochina (R18–R19), Myanmar
235 (R20), Sulawesi (R06), and northern Papua (R08), also exhibit relatively weak contributions from q ,
236 suggesting limited amplification by atmospheric moistening. In contrast, substantial humidity-driven
237 contributions are evident over the western Philippines (R01) and lowland monsoon regions of
238 Mainland SEA, including southern Vietnam (R17), Cambodia (R16), southern Thailand (R15), and
239 western Malaysia (R13) (Fig. 2c). These sub-regions stand out in Fig. 4a as exhibiting near-
240 comparable contributions from T and q , highlighting the important amplifying role of atmospheric



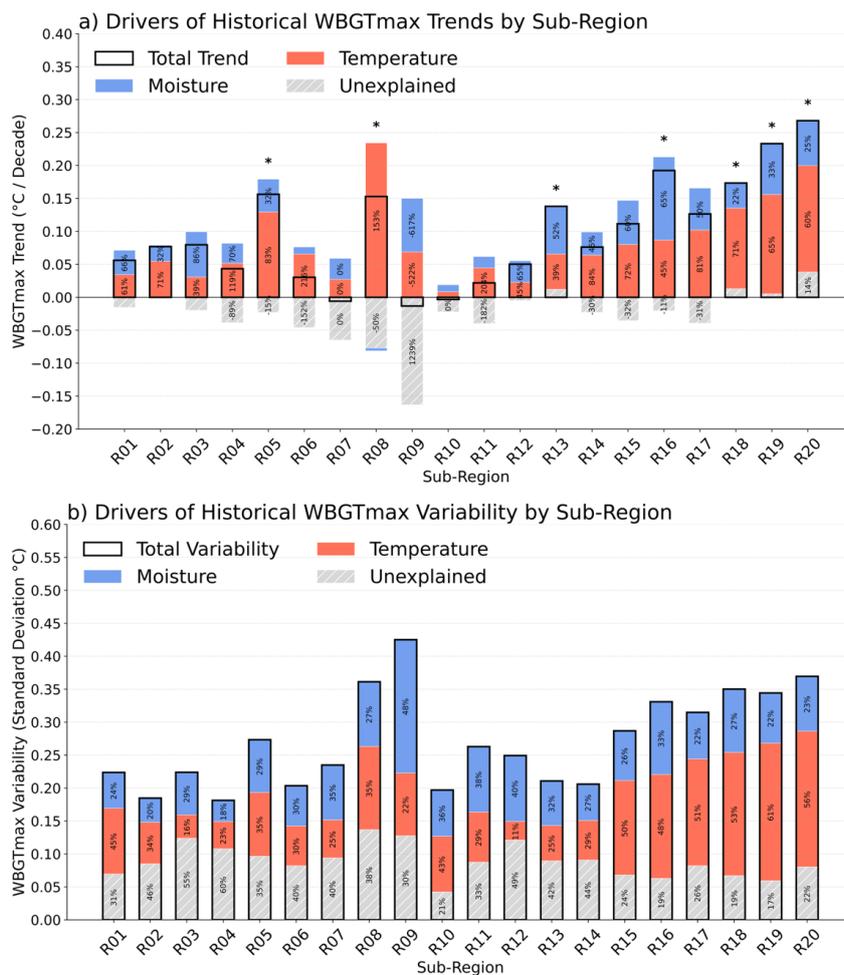
241 moistening. In these monsoon-affected environments, increasing q suppresses evaporative cooling
242 and enhances $WBGT_{max}$, consistent with a moist-heat intensification mechanism. Comparable q -
243 driven contributions are also found over northern Borneo (R03) and southern Papua (R09), although
244 the associated trends are generally not statistically significant.

245 The unexplained (residual) component (Fig. 2d) is spatially heterogeneous but typically smaller
246 than the thermodynamic terms and often negative, suggesting a modest damping relative to the
247 combined thermodynamic effects of T and q .



248

249 **Figure 3:** Spatial distribution of historical interannual variability in warm-season (April–October) daily
250 maximum wet-bulb globe temperature ($WBGT_{max}$) over Southeast Asia during 1985–2014. (a) Interannual
251 variability, quantified by its standard deviation, of $WBGT_{max}$ (°C). (b–d) Relative contributions (%) of near-
252 surface air temperature (T), specific humidity (q), and the unexplained (residual) component to the $WBGT_{max}$
253 interannual variability.



254

255 **Figure 4:** Area-averaged attribution of warm-season (April–October) *WBGTmax* historical trends and
 256 interannual variability across 20 Southeast Asia sub-regions (R01–R20) during 1985–2014. Black outlines
 257 indicate the total trends or variance. (a) Contributions of air temperature (T ; red), specific humidity (q ; blue),
 258 and the unexplained (hatched grey) term to sub-regional *WBGTmax* trends ($^{\circ}\text{C decade}^{-1}$). Asterisks denote
 259 sub-regions with statistically significant trends at $p < 0.05$. (b) Fractional contributions (%) of T , q , and the
 260 unexplained term to total *WBGTmax* interannual variance ($^{\circ}\text{C}$).

261 Figure 3a presents the interannual variability of *WBGTmax*, quantified by its standard deviation,
 262 during 1985–2014. Larger variability of about 0.3–0.4 $^{\circ}\text{C}$ or even higher is found over much of
 263 Mainland SEA and parts of the Maritime Continent, with pronounced maxima over northern
 264 Indochina and easternmost Indonesia in the Southern Hemisphere. In contrast, equatorial oceanic



265 regions and the southern Philippines display weaker fluctuations, generally below 0.3 °C, consistent
266 with strong ocean–atmosphere coupling and muted year-to-year temperature fluctuations. This
267 pattern indicates greater sensitivity of continental and highland areas to year-to-year climate
268 variations compared with maritime environments.

269 The attribution of interannual *WBGT_{max}* variability (Figs. 3b–d), together with the
270 corresponding variability in *T* and *q* (Fig. S2), shows a clear though not absolute mainland–maritime
271 contrast. Over areas north of approximately 10°N including Mainland SEA and northern Philippines,
272 variability in *T* explains a large fraction of *WBGT_{max}* variability (Fig. 3b). This reflects a
273 temperature-limited regime in which year-to-year variations in large-scale circulation, monsoon
274 strength, and episodic cold-air intrusions efficiently translate into WBGT anomalies. In contrast,
275 across much of the Maritime Continent (R05–R12), including Sumatra, Borneo, and western
276 Indonesia, *q* becomes an important, though not dominant, contributor (Fig. 3c). In these persistently
277 warm environments, relatively weak interannual variability in *T* allows fluctuations in *q* associated
278 with circulation anomalies associated with ENSO and the Madden–Julian Oscillation to exert a
279 strong influence on heat-stress variability.

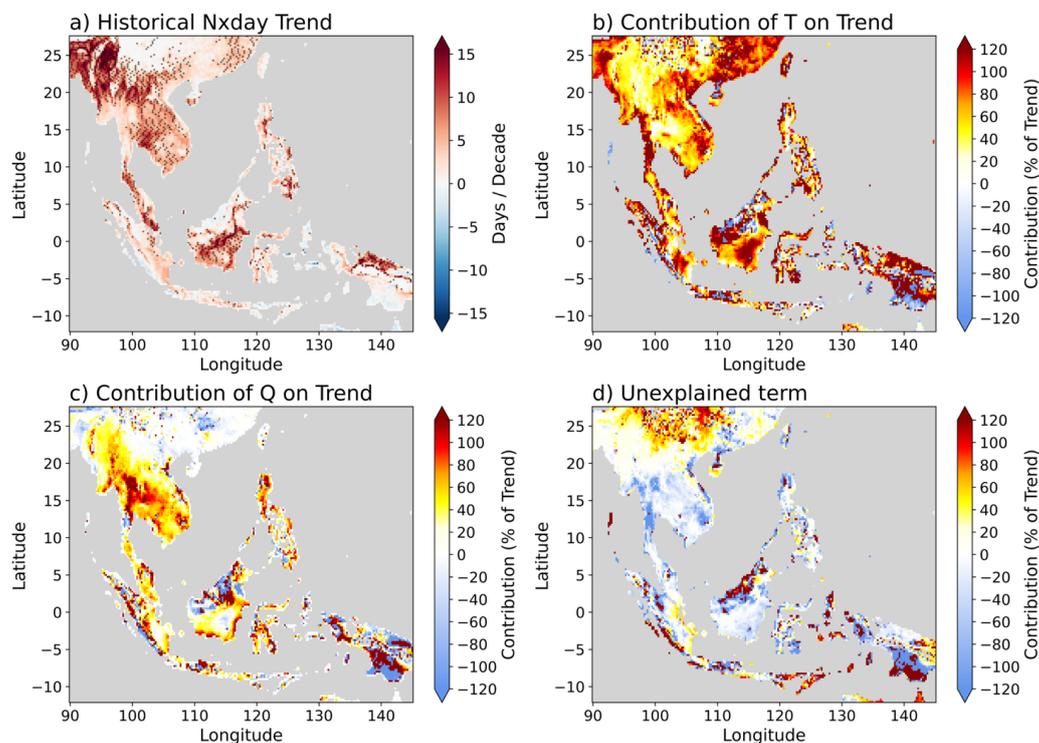
280 Notably, the unexplained (residual) component (Fig. 3d) accounts for a substantial fraction of
281 total interannual variance over large parts of SEA, particularly over the eastern Philippines (R02) and
282 much of the Maritime Continent (R03–R14). In many sub-regions, the unexplained component
283 equals or exceeds the individual *T* or *q* contributions. This indicates that nonlinear temperature–
284 humidity interactions and compound anomalies play a major role in shaping interannual extreme heat
285 stress, beyond what is captured by a linear additive framework. Together, these patterns identify two
286 distinct variability regimes: a temperature-controlled regime over monsoon-influenced continental
287 regions, and an interaction-dominated regime over the Maritime Continent, where persistently high
288 background humidity makes *WBGT_{max}* variability highly sensitive to compound thermodynamic
289 anomalies rather than to variability in a single driver. The regional summary (Fig. 4b) highlights



290 these regimes. This shows that while T dominates variability in Mainland SEA sub-regions,
 291 Maritime Continent regions exhibit a more even partitioning among T , q , and the unexplained
 292 component, with the latter exceeding 40% in several sub-regions (e.g., R03–R04, R12–R14).

293 Comparing this with the trend attribution reveals an important contrast: long-term trends in
 294 $WBGT_{max}$ reflect relatively smooth, additive thermodynamic forcing, whereas interannual
 295 variability arises primarily from episodic compounding of temperature and moisture anomalies.

296 3.2 Drivers of historical (1985–2014) heat stress frequency



297

298 **Figure 5:** Spatial distribution of historical long-term trends in warm-season (April–October) annual number
 299 of extreme heat-stress days (N_{xday}) over Southeast Asia during 1985–2014. (a) Linear trend in N_{xday} (days
 300 decade⁻¹); stippling denotes statistical significance at $p < 0.05$. (b–d) Relative contributions (%) of air
 301 temperature (T), specific humidity (q), and the unexplained (residual) component to N_{xday} trends.



302 Building on the analysis of *WBGTmax* intensity, we next assess changes in the frequency of
303 extreme heat stress using *Nxday*. The spatial distribution and dominant controls closely resemble
304 those identified for *WBGTmax*, indicating common thermodynamic drivers. During the historical
305 period (1985–2014), *Nxday* exhibits widespread increasing trends across SEA (Fig. 5a). Statistically
306 significant trends are concentrated over mainland regions, western Malaysia, and parts of the
307 Maritime Continent. In several locations, particularly northwestern Indochina and high-elevation
308 areas, increases reach roughly 8–10 days decade⁻¹, indicating a rapid rise in the occurrence of
309 extreme heat-stress conditions.

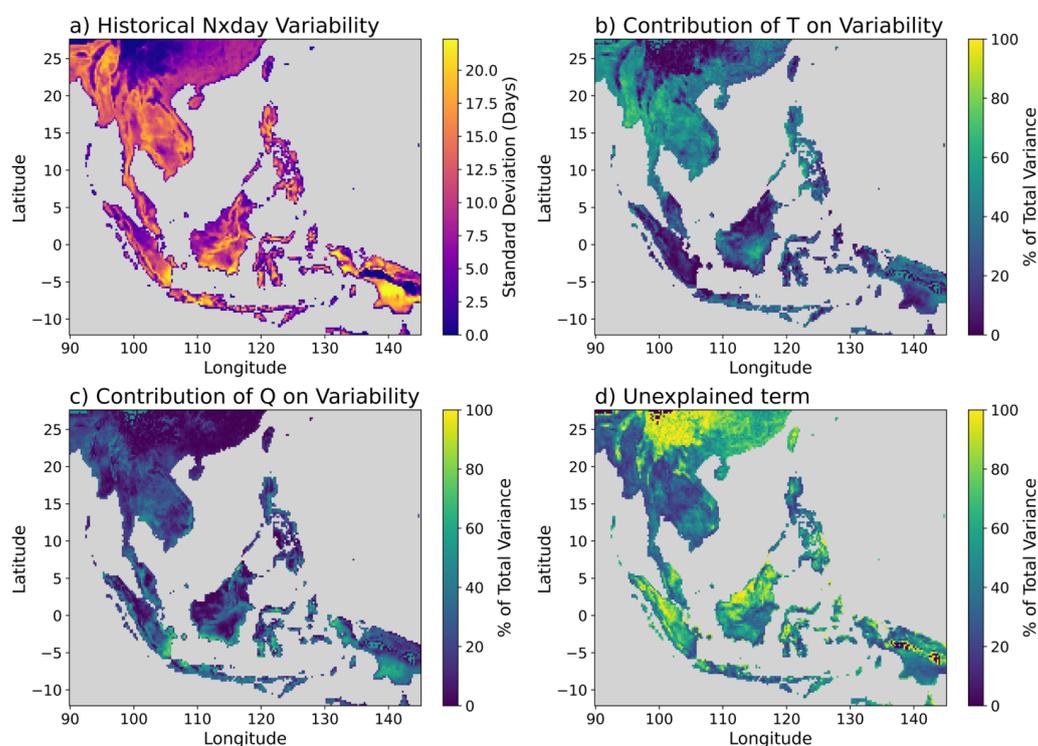
310 The decomposition points to a dominant, though not exclusive, role of *T* (Fig. 5b). Over large
311 parts of the domain, warming explains most of the increase in threshold exceedances, reflecting the
312 strong sensitivity of *Nxday* to shifts in the background climate. Moisture contributions (Fig. 5c) are
313 more evident in monsoon-influenced lowlands, where higher humidity raises baseline WBGT and
314 increases the probability of surpassing the threshold, consistent with the behavior found for
315 *WBGTmax*. In contrast, many Maritime Continent regions display weaker and less coherent
316 contributions from *q*, consistent with smaller mean changes and stronger modulation by variability.
317 The unexplained (residual) component (Fig. 5d) is spatially heterogeneous and generally negative for
318 long-term trends, implying a modest damping of the combined thermodynamic effect.

319 The sub-regional synthesis (Fig. 5a) corroborates this picture. Northern Mainland SEA sub-
320 regions (R18–R20) show robust, temperature-dominated increases in *Nxday*, while *q* plays a
321 secondary but visible role in lowland monsoon regions such as R13 and R15–R17, and in several
322 maritime regions (e.g., R05 and R07). Overall, maritime areas exhibit smaller and less consistently
323 significant trends, consistent with weaker warming and stronger modulation by moisture variability.

324 Interannual variability in *Nxday* (Fig. 6a) is substantial, often exceeding 10 days over both
325 Mainland SEA and Maritime Continent. Compared with *WBGTmax*, however, the attribution reveals
326 a stronger dominance of the unexplained component in variability in *Nxday*. Although variations in *T*



327 and q contribute locally (Figs. 6b–c), neither variable alone accounts for most of the variance across
328 large portions of SEA. Instead, the unexplained term (Fig. 6d) explains a major share of the total
329 variability, indicating an essential role for nonlinear threshold behavior and compound anomalies.



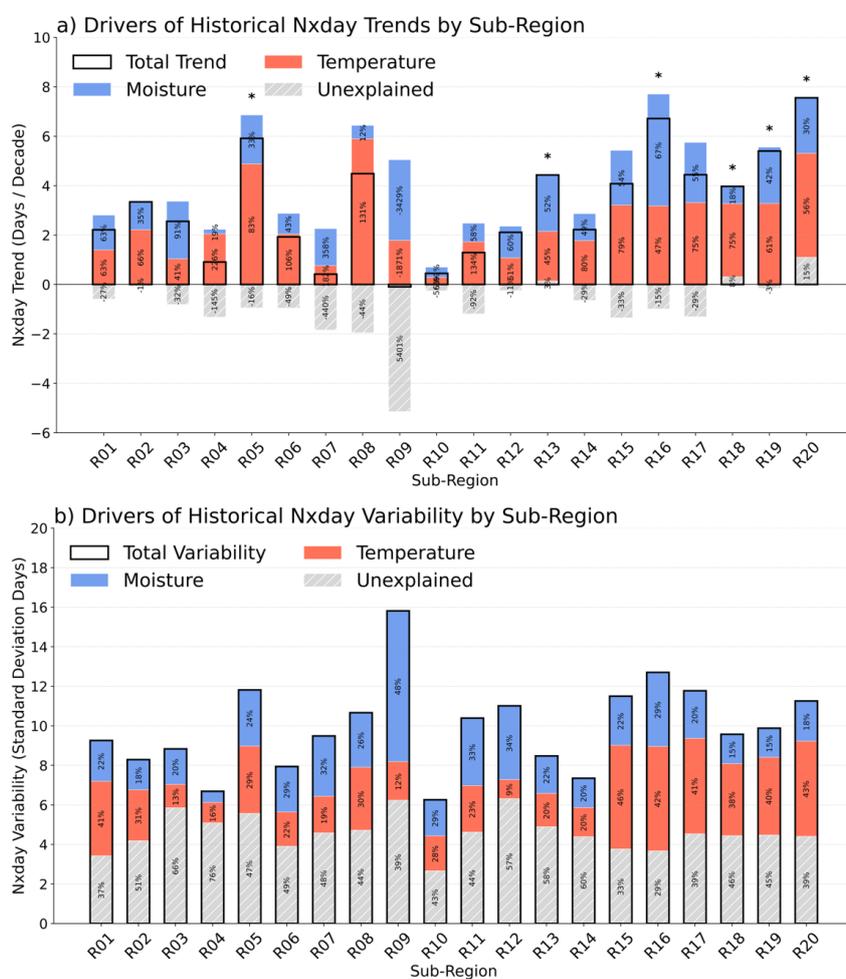
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331 **Figure 6:** Spatial distribution of historical interannual variability in warm-season (April–October) annual
332 number of extreme heat-stress days ($Nxday$) over Southeast Asia during 1985–2014. (a) Interannual
333 variability, quantified by its standard deviation, of $Nxday$ (days). (b–d) Relative contributions (%) of near-
334 surface air temperature (T), specific humidity (q), and the unexplained (residual) component to the $Nxday$
335 interannual variability.

336 The physical interpretation of this residual differs by region. In monsoon-dominated areas, year-
337 to-year, year-to-year variability in $Nxday$ often reflects episodic warm and humid seasons that shift a
338 large part of the distribution above the critical threshold. Meanwhile, over the Maritime Continent,
339 where background humidity is persistently high, relatively small concurrent changes in T – q can
340 produce disproportionately large changes in exceedance frequency. This regime dependence is



341 confirmed by the sub-regional synthesis (Fig. 7b), in which the unexplained component typically
 342 accounts for 40–60% of total *Nxday* variability. The result underscores that compound
 343 thermodynamic anomalies, rather than fluctuations in a single driver, govern much of the variability
 344 in extreme heat-stress frequency.

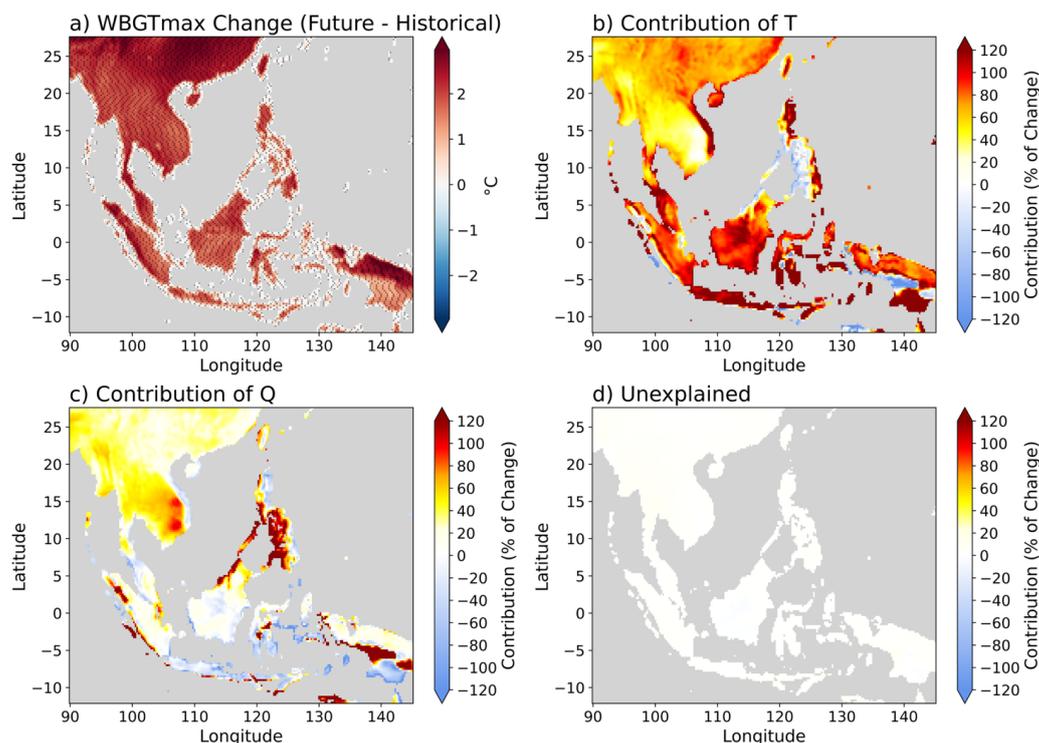


345

346 **Figure 7:** Area-averaged attribution of warm-season (April–October) *Nxday* historical trends and interannual
 347 variability across 20 Southeast Asia sub-regions (R01–R20) during 1985–2014. Black outlines indicate the
 348 total trends or variance. (a) Contributions of air temperature (T ; red), specific humidity (q ; blue), and the
 349 unexplained (residual; hatched grey) term to sub-regional *Nxday* trends (days decade⁻¹). Asterisks denote sub-
 350 regions with statistically significant trends at $p < 0.05$. (b) Fractional contributions (%) of T , q , and the
 351 unexplained component to total *Nxday* interannual variance (days).



352 **3.2 Drivers of future (2071–2100) minus (1985–2014) changes**



353

354 **Figure 8:** Spatial patterns of projected changes in warm-season (April–October) daily maximum wet-bulb
 355 globe temperature (*WBGTmax*) over Southeast Asia between the historical (1985–2014) and the future (2071–
 356 2100) periods under the SSP5-8.5 scenario. (a) The total change (future minus historical periods); stippling
 357 denotes statistical significance at $p < 0.05$. (b), (c) and (d) Relative contributions (%) of air temperature (*T*),
 358 specific humidity (*q*) and unexplained (residual) component, respectively, to projected *WBGTmax* changes.

359 Future projections under SSP5–8.5, relative to the historical climatology (Figs. S3 and S4), show
 360 large and spatially coherent increases in warm-season *WBGTmax* across SEA (Fig. 8a). Regional-
 361 mean increases typically exceed approximately 1.5–2 °C and are locally higher over mainland areas
 362 and parts of the eastern Maritime Continent. The widespread and spatially consistent nature of the
 363 projected warming indicates a basin-scale upward shift in the *WBGTmax* distribution, rather than
 364 localized or region-specific intensification. This coherence points to large-scale thermodynamic
 365 forcing as the primary driver of future heat-stress intensification.



366 The attribution indicates that T remains the dominant driver to future increases in $WBGT_{max}$,
367 particularly over the Maritime Continent (Fig. 8b). However, unlike the historical period, q now
368 contributes positively and often substantially throughout the region (Fig. 8c). Over monsoon-
369 influenced regions, including the Philippines and lowland Mainland SEA, the contribution from q
370 frequently accounts for approximately 30–50% of the projected $WBGT_{max}$ increase. This
371 widespread amplification reflects concurrent warming and moistening expected in a warmer climate
372 and signals a transition toward more compound warm–humid conditions, consistent with Clausius–
373 Clapeyron scaling and projected increases in both T and q under SSP5–8.5 (Fig. S5).

374 The unexplained (residual) contribution is negligible over most areas (Fig. 8d), indicating that
375 future $WBGT_{max}$ changes well explained by first-order thermodynamic adjustments. The sub-
376 regional synthesis (Fig. 10a) reinforces this finding: All sub-regions exhibiting statistically
377 significant increases in $WBGT_{max}$, with T dominating across most Maritime Continent sub-regions
378 (R03–R14), while monsoon-influenced sub-regions of Mainland SEA and the Philippines (R01–R02,
379 R15–R20) show more comparable roles from T and q . Together, these results highlight a growing
380 control of moist enthalpy on future heat stress, rather than dry-bulb warming alone.

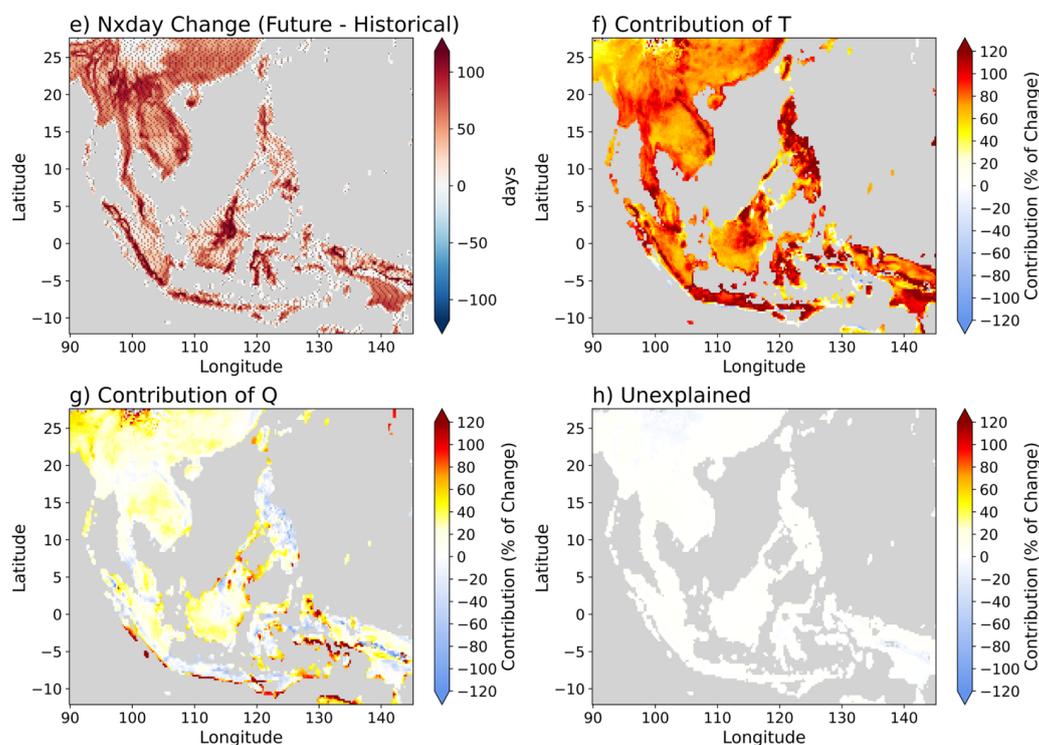
381 The frequency of extreme heat stress N_{xday} also increases dramatically by the end of the century
382 under SSP5–8.5 (Fig. 9a) compared with the historical climatology (Figs. S3 and S4). Many regions
383 experience rises of several tens of days per year, with some mainland and maritime locations
384 exceeding ~ 50 – 60 days $year^{-1}$. The strong spatial similarity with $WBGT_{max}$ changes indicates that
385 higher background heat stress translates directly into more frequent threshold exceedances.

386 Temperature again provides the dominant contribution (Fig. 9b). Compared with $WBGT_{max}$, the
387 contribution from q is systematically weaker (Fig. 9c). While background moistening elevates
388 baseline WBGT, the frequency of threshold exceedance is primarily governed by shifts in mean T
389 relative to a fixed extreme heat-stress threshold. The unexplained (residual) term remains minimal



390 (Fig. 9d), underscoring that projected *Nxday* changes are largely governed by straightforward

391 thermodynamic forcing.



392

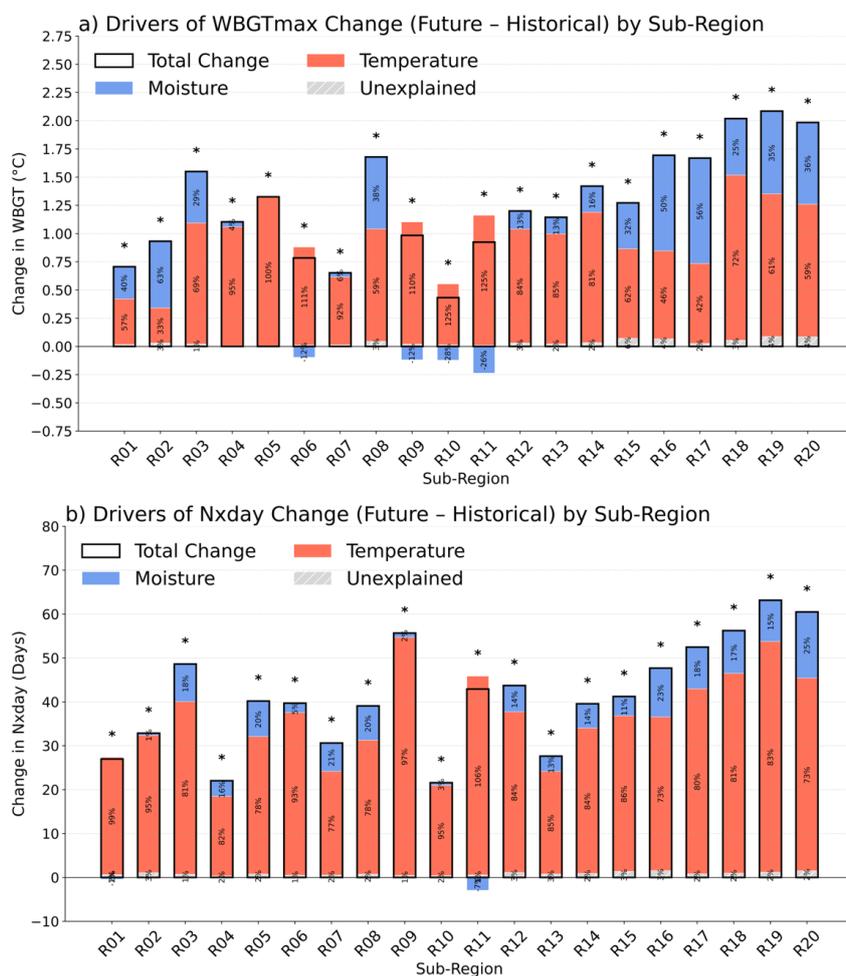
393 **Figure 9:** Spatial patterns of projected changes in warm-season (April–October) annual number of extreme
 394 heat-stress days (*Nxday*) over Southeast Asia between the historical (1985–2014) and the future (2071–2100)
 395 periods under the SSP5-8.5 scenario. (a) The total change (future minus historical periods); stippling denotes
 396 statistical significance at $p < 0.05$. (b), (c) and (d) Relative contributions (%) of air temperature (T), specific
 397 humidity (q) and unexplained (residual) component, respectively, to projected *Nxday* changes.

398 Despite broadly similar spatial patterns, *Nxday* exhibits a stronger amplification of future
 399 changes than *WBGTmax* owing to its nonlinear response to mean-state warming. Even modest
 400 increases in mean WBGT can produce large rises in the number of exceedance days, particularly
 401 in regions already close to critical heat-stress thresholds. This sensitivity is confirmed in the sub-
 402 regional synthesis (Fig. 10b), where T explains approximately 70–90% of projected *Nxday* increases,



403 while q provides a secondary but regionally important enhancement, particularly in monsoon-

404 influenced and coastal environments.



405

406 **Figure 10:** Area-averaged attribution of future minus historical changes in warm-season (April–October) (a)
 407 *WBGTmax* and (b) *Nxday* across 20 Southeast Asia sub-regions (R01–R20). Black outlines indicate the total
 408 change between 1985–2014 and 2071–2100. Colored bars represent contributions from temperature (T ; red),
 409 specific humidity (q ; blue), and the unexplained component (hatched grey). Asterisks denote sub-regions with
 410 statistically significant total changes at $p < 0.05$.



411 **4. Discussions and Conclusions**

412 This study presents a unified thermodynamic attribution of warm-season (April–October) heat
413 stress across Southeast Asia (SEA) by separating the contributions of near-surface air temperature
414 (T), specific humidity (q), and non-linear unexplained (residual) processes across multiple
415 timescales.

416 During the historical period (1985–2014), increases in both heat-stress intensity ($WBGT_{max}$)
417 and frequency ($Nxday$) are driven primarily by rising T , particularly over the Maritime Continent and
418 elevated terrain. Nevertheless, q already exerts an important amplifying influence in monsoon-
419 affected lowlands, including Indochina, the Philippines, and parts of the Malay Peninsula. These
420 results demonstrate that contemporary heat stress in SEA cannot be understood from temperature
421 change alone.

422 Future projections (2071–2100) under SSP5-8.5 reveal a systematic strengthening of compound
423 warm–humid conditions by the end of the 21st century. Although T remains the leading contributor
424 to projected increases in both $WBGT_{max}$ and $Nxday$, q becomes uniformly positive and often
425 comparable in magnitude, especially in monsoon-influenced regions. This response is consistent with
426 thermodynamic expectations of enhanced atmospheric water-holding capacity in a warmer climate
427 and implies a continued decline in evaporative cooling efficiency. Consequently, late-century heat
428 stress is increasingly governed by elevated moist enthalpy rather than dry-bulb warming alone.

429 A further distinction arises between long-term trends and interannual variability. Historical
430 trends and future mean-state shifts explained by relatively smooth thermodynamic adjustments,
431 whereas year-to-year variability, particularly over the Maritime Continent, is strongly shaped by non-
432 linear interactions between T and q . This behavior is most pronounced for exceedance-based metrics
433 such as $Nxday$, where small co-occurring anomalies can trigger large changes in frequency. Hence,
434 while gradual warming provides a predictable background intensification, seasonal extremes remain



435 strongly influenced by compound processes whose importance varies between continental monsoon
436 and maritime regimes.

437 The projected intensification of extreme heat stress has important implications for hazard
438 management and adaptation across SEA. More frequent exceedances of extreme WBGT thresholds
439 point to increasing risks for outdoor labor and vulnerable populations, particularly in regions with
440 high population density and limited adaptive capacity. Traditional protective measures such as rest
441 scheduling, hydration, and workload modification will remain necessary but may become less
442 effective as humidity rises (Morabito et al., 2019). Early-warning systems and risk frameworks that
443 incorporate both T and q are therefore likely to outperform approaches based solely on T . At the
444 same time, expanding demand for mechanical cooling raises challenges for energy infrastructure and
445 emissions, emphasizing the need for coordinated strategies linking climate risk, urban planning, and
446 public health.

447 Several limitations warrant consideration. This study does not explicitly quantify labor
448 productivity losses or economic impacts, which are known to be highly sensitive to WBGT threshold
449 exceedances. In addition, urban heat-island (UHI) effects are not resolved at the sub-regional scale
450 and may further amplify heat stress in major metropolitan areas. The projections also assume limited
451 adaptation and should therefore be interpreted as upper-bound estimates under the assumed
452 emissions scenario (Russo et al. 2019). Finally, while wind and radiation are included within WBGT
453 calculations, their individual contributions are not separately attributed, and evolving circulation
454 variability could further shape future extremes. Addressing these issues represents a priority for
455 future work.

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463 **Data Availability** Outputs from CMIP6 global climate models are accessed via <https://esgf->

464 node.ipsl.upmc.fr/projects/cmip6-ipsl/, while the ECMWF ERA5 was downloaded from

465 <https://www.ecmwf.int/en/forecasts/dataset/ecmwf-reanalysis-v5>. The RegCM downscaling outputs

466 and the WBGT datasets generated in this study are available from the corresponding author upon

467 reasonable request.

468 **Code availability** Codes used in this study are available upon reasonable requests from the

469 corresponding author.

470 **Author contributions** The author conceived the study, performed the analysis, and wrote the

471 manuscript.

472 **Declarations**

473 **Conflict of interest** The authors declare no competing interests.

474 **Ethical approval** Ethical approval is not applicable to this article.

475 **References**

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