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1 **Anthropogenic influences on the persistent night-time heat wave in**
2 **summer 2018 over North-East China**

3

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30 **Capsule summary**

31 Persistent night-time heat waves like the event of summer 2018 in North-East China
32 are extremely rare (about a one-in-a-500-year) in the natural world, but now have
33 become about a one-in-60-year event with anthropogenic warming.

34

35 **Introduction**

36 In the summer of 2018, North-East China was affected by an unprecedentedly long
37 and intense heat wave. The China Meteorological Administration issued 33 days of
38 consecutive "high temperature-alert" warnings from 14th July to 15th August in 2018.
39 Record-breaking hot minimum temperatures were observed in a large area of North-
40 East China (34°N-55°N, 105°E-135°E) with a stable spatial pattern on timescales of
41 20-40 days (Fig. 1a, S1). Further, minimum temperatures were more extreme, with a
42 much larger record-breaking area than maximum temperatures (Fig. S1). On the 30th
43 July, the number of heat-related hospitalization admissions broke the historical record
44 in Shenyang, a large city in Northeast China
45 (<http://news.lnd.com.cn/system/2018/08/01/000008645.shtml>). The aquaculture
46 industry in Liaoning Province suffered from economic loss of 6.87 billion RMB
47 (<https://www.zhonghongwang.com/show-256-103674-1.html>). Thus, this
48 unprecedented persistent and extreme heat wave event led to severe impacts,
49 including increased human morbidity and mortality, reduced agriculture productivity
50 and increased strain on power systems and water supplies (CMA 2018).

51 Anthropogenic warming has been shown to drive recent record-breaking heat and
52 summer extremes in different regions of the world (Hansen et al., 2012; Lewis et al.,
53 2016). Previous heat event attribution studies in China usually considered seasonal
54 mean and maximum temperature covering a fixed period (e.g., Sun et al., 2016; Ma et
55 al. 2017), with few studies focusing on consecutive minimum temperatures when
56 there is the strongest signal in summer. Daily minimum temperature allows people
57 and ecosystems to recover from thermal stresses experienced during the previous day

58 (Schwartz et al. 2005) and is a strong predictor for human morbidity and mortality
59 (Laaidi et al. 2012; Madrigano et al. 2015; Murage et al. 2017). Previous studies show
60 that anthropogenic influences, including anthropogenic emission of greenhouse gases
61 and urbanization leading to the expansion of urban heat islands, has contributed to
62 higher summer minimum temperatures in Eastern China (Sun et al. 2014; Wang et al.
63 2017). Hence, this study aims to investigate whether anthropogenic influences have
64 increased the frequency of the monthly-time scale heat waves like the summer 2018
65 event over Northeast China.

66 Data and Methods

67 Observations

68 We used observed daily minimum temperatures at about 2400 meteorological stations
69 over China for the period 1961-2018. These were quality-controlled and homogenized
70 by National Meteorological Information Center of China (Ren et al. 2012). We also
71 used daily atmospheric circulation field data from ERA-interim (Dee et al, 2010),
72 including geopotential height, horizontal wind and specific humidity.

73 Model

74 Simulations from the atmosphere model HadGEM3-A-N216 at a horizontal resolution
75 of $0.56^{\circ} \times 0.83^{\circ}$ were used in this study (Christidis et al. 2013; Ciavarella et al. 2018).

76 Three ensembles were used:

- 77 • An 15-member ensemble of simulations for period 1960-2013, in which the
78 model is forced by observed sea surface temperatures (SST) and sea ice
79 concentrations (SIC) from HadISST (Rayner et al. 2003), and a
80 comprehensive package of historical anthropogenic atmospheric forcing
81 (Historical).
- 82 • A 525-member ensemble of simulations for 2018 only (counterfactual world),
83 driven with preindustrial atmospheric forcing and the anthropogenic
84 contribution removed from SST and SIC (HistoricalNatExt. See SI for details).

- 85 • A second 525-member ensemble of simulations (factual world), driven with
86 the same as “Historical” but for 2018 only (HistoricalExt).

87 Event Definition

88 An index of 30-day moving average of daily minimum temperature anomalies
89 (TNx30) over the study area was defined. First, to remove the seasonal cycle at each
90 station, daily minimum temperature (Tmin) anomalies relative to the daily 1961-2013
91 climatology were computed in each calendar day. We then gridded these anomalies to
92 the $0.56^{\circ} \times 0.83^{\circ}$ model resolution by averaging all station anomalies in each grid box.
93 The gridded Tmin anomalies were then area averaged over North-East China (34°N -
94 55°N , 105°E - 135°E), and finally used to calculate the hottest 30-day running mean
95 Tmin in summer (June-August) for each year. For model datasets, the same process
96 was followed except rather than gridding the simulations, grid-points were used only
97 if they were land and there were observations in the grid-box. The anomaly of the
98 index ‘TNx30’ in summer was used in this study and the value of 2018 event (2.15°C)
99 was chosen as the threshold.

100 To estimate probabilities, generalized extreme value (GEV) distributions were fitted
101 to both simulated and observed data. A two-side Kolmogorov-Smirnoff test was then
102 applied to test if the distributions of the observations and historical simulations from
103 1961 to 2013 are from the same population. To estimate the anthropogenic
104 contribution to the heat wave event like 2018 summer in North-East China, the risk
105 ratio (National Academies of Sciences, Engineering, and Medicine 2016) defined as
106 $P(\text{HistoricalExt})/P(\text{HistoricalNatExt})$ was calculated, where $P(\text{HistoricalExt})$ and
107 $P(\text{HistoricalNatExt})$ are the probability of the event in factual and counterfactual
108 world, respectively. We bootstrapped with replacement 1000 times to generate PDFs,
109 and then computed 1000 risk ratios and then used that to compute uncertainties in risk
110 ratios.

111 3. Results

112 3.1 Observed events and model performance

113 In summer 2018, the regional average of TNx30 over North-East China was 3
114 standard deviations of inter-annual variability above the 1961-2013 climatology and
115 the highest on record since at least 1961 (Fig. 1b). This heat wave was accompanied
116 by positive geopotential height anomalies over Northeast Asia, induced by the
117 unprecedented northward shift of the Western Pacific subtropical high (Fig. 1c; Liu et
118 al. 2019). At low-levels (850hPa), anomalous northwestward moisture transportation
119 from the warm Bohai Sea resulted in increased specific humidity (Fig. 1c) and
120 consequently contributed to significant night-time warming.

121 Model performance was evaluated using the ensemble mean of the historical
122 ensemble, which reasonably reproduced the time series of TNx30 anomalies, with a
123 correlation coefficient of 0.70 (Fig. 2a). This means that forcing, SST and SIC
124 variations explains about half of the observed variance in TNx30. The distributions
125 of observed and simulated TNx30 for summers during 1961-2013 are also statistically
126 indistinguishable based on K-S test ($p=0.80$, Fig. 2b). Such good performance of the
127 HadGEM3-A-N216 simulations provides the basis for further attribution analysis.

128

129 3.2 Anthropogenic impact on the risk of heat waves

130 There is a shift of the PDF to warm anomalies from historicalNatExt to historicalExt
131 (Fig. 2c), indicating that anthropogenic influences have increased the probability of
132 heat wave events. Since the magnitude of this event lies at the far warm-end tail of
133 PDF, events like 2018 are very rare in the counterfactual world. Only one member in
134 525-member historicalNatExt ensemble exceeds the 2018 threshold. By contrast, the
135 estimated probability is 0.02 in factual world. These indicate that the 2018-like night-
136 time heat event is extremely rare without anthropogenic warming. The estimated
137 return period of heat wave events hotter than 2018 is about one-in-60-years with

138 anthropogenic warming, with 5th-95th percentile uncertainty ranges of 43-116 years
139 (Fig. 2d). A second threshold was also selected, defined as the second-most extreme
140 year (2017, with an anomaly of 1.55°C). For this threshold, the heatwave is 57 times
141 more frequent in the factual world ($P(\text{HistoricalExt}) = 0.17$) than the counterfactual
142 world ($P(\text{historicalNatExt}) = 0.003$), which confirms the role of anthropogenic
143 warming in these heat events. In terms of return period, for a one-in-10-year event,
144 the magnitude of TNx30 in 2018 is estimated to be 1.7°C (1.7-1.8, 5th-95th) in the
145 factual world, and 0.8 °C (0.8-0.9 5th-95th) in the counterfactual world (Fig. 2d). For a
146 one-in-50-year event, the estimated magnitude of a heat event is 2.1°C (2.0-2.2) and
147 1.2°C (1.1-1.3) in the counterfactual and factual worlds (Fig. 2d). The change in
148 return level (0.9°C) between counterfactual and factual simulations is generally
149 consistent with mean warming, as the shift in mean state between with (0.95°C) and
150 without (-0.04°C) anthropogenic influence was 0.99°C (Fig. 2c). Besides, the
151 uncertainty range of return periods increases with the rarity of events. We repeat our
152 analyses by using different durations of either 20 day or 40 day moving average and
153 find similar risk ratio, suggesting the robustness of the results.

154 **4. Conclusions and discussion**

155 North-East China experienced a record-breaking night-time heat wave in 2018
156 summer. This kind of 30-day night-time heatwave was, a one-in-500 year or less,
157 event in the counterfactual world. Forced by anthropogenic forcing and the observed
158 2018 SSTs, it became a one-in-60-year event.

159 This unprecedentedly long-lasting night-time heat wave was also related to the
160 northwestward shift of West Pacific Subtropical High and the anomalous moisture
161 transportation from the warm ocean. Additionally, the configuration of anomalous
162 anticyclone at 500-hPa over Northeast Asia during this heat event was reproduced
163 with the six hottest simulations in study region (Fig S2), confirming the role of
164 abnormal high-pressure system in this heat wave event. As increased occurrence of
165 anticyclonic circulations in the mid-latitude has made a substantial contribution (one-
166 third to one-half) to the increased summertime temperature extremes over portions of

167 Eurasia since 1979 (Horton et al. 2015), this leaves further questions whether
168 anthropogenic warming has contributed to the heat waves like in summer 2018
169 through affecting the background circulation. This study used an atmospheric model
170 conditioned on the observed SSTs. Results are inevitably affected by uncertainty in
171 the representation on the SSTs in the counterfactual world, especially for severe
172 events with return period greater than 50 years (e.g., Sparrow et al. 2018), and more
173 work is necessary to better understand this uncertainty.

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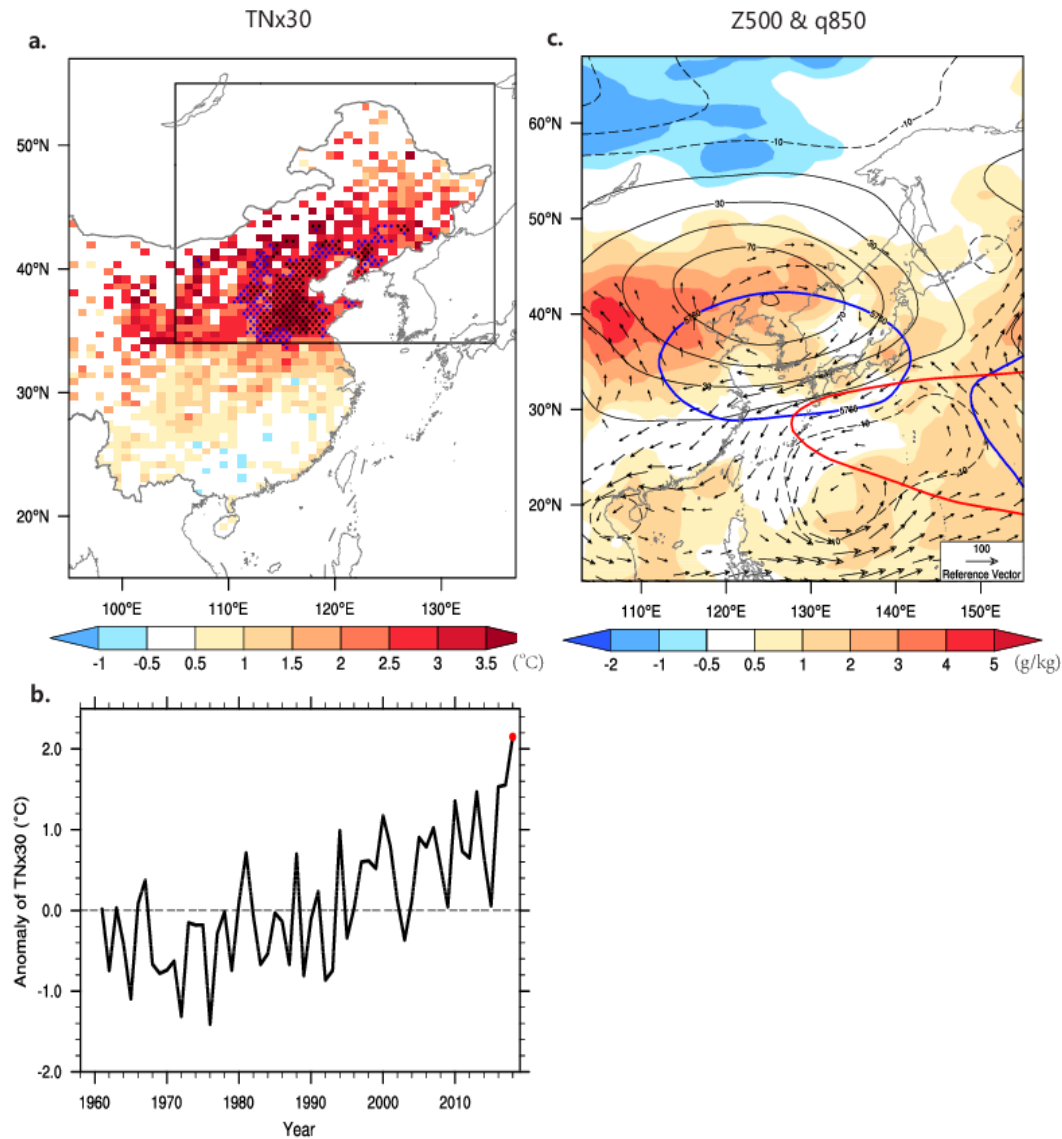
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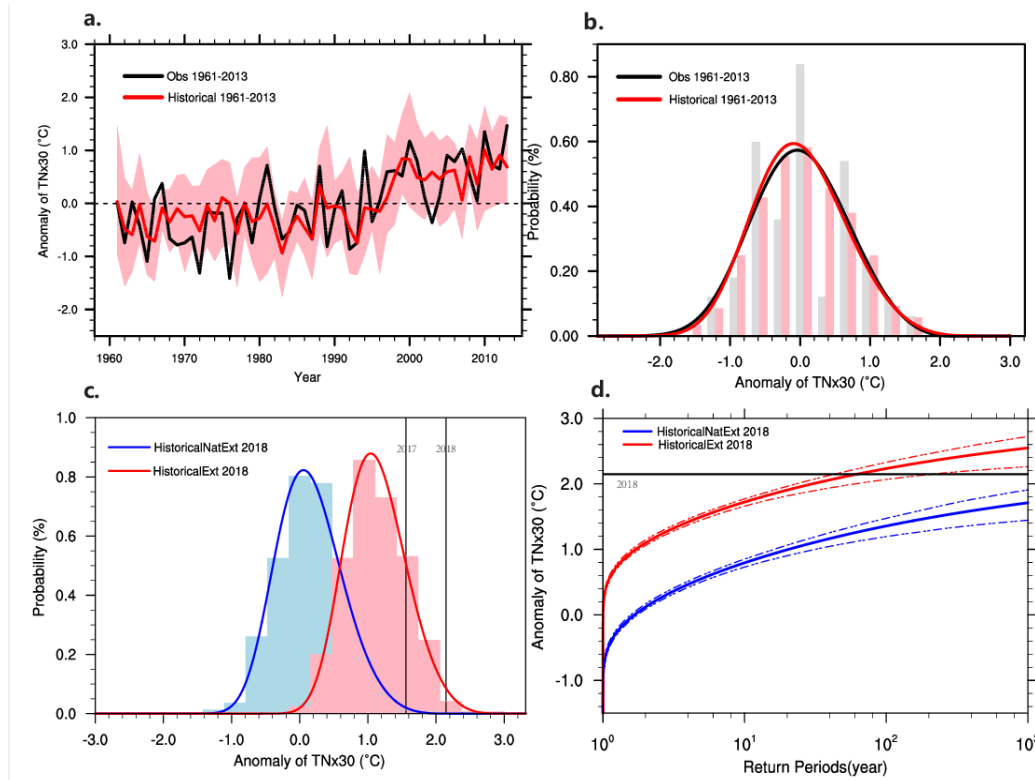
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270

271 **Fig1.** Observed characteristics of the heat wave in Northeast China during 12th July –
 272 10th August, with maximum consecutive 30-day Tmin anomalies in summer 2018
 273 (TNx30). (a) Spatial pattern of TNx30 (shading, unit:°C) relative to 1961-2013.
 274 Locations with record-breaking and second highest values since 1961 are shown with
 275 black and blue dots, respectively. (b) Time series of TNx30 anomalies over Northeast
 276 China (black rectangle shown in Fig1 a) from 1961-2018. (c) Circulation field from
 277 ERA-interim with specific humidity anomalies (shading, unit: g/kg) and 850-hPa
 278 moisture flux anomalies (vectors). The light black contours denote the 500-hPa
 279 geopotential height anomalies. 12th July – 10th August mean geopotential height
 280 (blue lines) and climatology (red lines) are also shown.



281

282 **Fig2.** (a) Time series of observed (black) and simulated ensemble mean (red) of
 283 TNx30 anomalies, with 15-member spread shown as light pink shading. (b)
 284 Histograms and GEV-fit probability density function of TNx30 anomalies in
 285 observations (black line and grey bars) and historical simulations (red line and light
 286 pink bars) from 1961-2013. (c) Histograms and GEV-fit probability density function
 287 (PDF) of TNx30 anomalies in 2018 summer with 525-member historicalExt (red) and
 288 historicalNatExt (blue) simulations. The black lines indicate the TNx30 thresholds of
 289 2018 and 2017. (d) same as (c) but for return periods. Dash-dot lines indicate the 5th-
 290 95th uncertainty range based on 1000 bootstrap resamples for All-forcing (red) and
 291 natural-forcing (blue).