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To cite this article: Eun-Soon Im et al 2019 Environ. Res. Lett. 14 094020

View the article online for updates and enhancements.
Environmental Research Letters

LETTER

2018 summer extreme temperatures in South Korea and their intensification under 3 °C global warming

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Keywords: fine-scale climate projections, dynamical downscaling, 2018 Korean heatwaves, regional impacts of global warming, intensified future heat stress

Supplementary material for this article is available online

Abstract

With the acceleration in global warming, extreme hot temperatures have emerged as one of the most prominent risks. In this study, we characterize the unprecedented extreme temperatures that occurred in Korea in summer 2018, and attempt to explain how this locally observed extreme event can be interpreted in the context of 2 °C and 3 °C global warming above the pre-industrial level. To better resolve geographically diverse climate features and enhance confidence in future changes, three global projections are dynamically downscaled using three regional climate models that are customized over Korea and the systematic biases are statistically corrected using quantile mapping. In July and August 2018, abnormally high maximum temperatures (Tmax) were observed over the entire territory of South Korea. Beyond the increase of mean value, Tmax at individual stations departed significantly from the typical Gaussian distribution of climatological Tmax due to the dramatic changes in the extent and shape of upper tails. The distinct behaviors of Tmax that appeared in 2018 largely represent the statistical analog of the distribution pattern expected under 3 °C global warming based on fine-scale climate projections. This implies that statistically extremely rare events like that of summer 2018 will become increasingly normal if global average temperature is allowed to increase by 3 °C. More importantly, the extreme heat stress measured by the wet-bulb globe temperature is projected to intensify the risks to a level never before seen in contemporary climate. This study is timely and relevant to the need to identify how the globally aggregated warming target temperature can be disaggregated into regional impacts.

1. Introduction

The United Nations Conference on Climate Change (COP21), which was held in Paris in December 2015, reached the consensus that significant efforts need to be made on the reduction of emissions to avoid disastrous consequences attributable to global warming. As a result, the world’s governments agreed to keep the increase in global average temperature to well below 2 °C above pre-industrial levels and to ambitiously aim to limit the increase to 1.5 °C (UNFCCC 2015). Setting out a clear target such as 2 °C warming seems to be an effective way to promote urgency and pursue global actions. On the other hand, it may give the public the wrong perception that the threshold of 2 °C can be regarded as a universally applicable goal and, therefore, that anthropogenic interference with the climate system will remain acceptable until global warming has reached 2 °C. However, any change in global average temperature is not linearly disaggregated into regional and local impacts in a straightforward manner (Knutti et al 2016,
Harrington et al. 2018). In particular, the changes in intensity and frequency of extreme events in response to global average temperature are highly nonlinear at regional scales (Ruff and Neelin 2012).

In summer 2018, South Korea experienced unprecedented extreme hot temperatures over its entire territory. The Korean Meteorological Administration (KMA) reported that record-breaking daily Tmax and consecutive hot days were repeatedly observed in 2018 (KMA 2019). For example, Tmax in Seoul rose to 39.6 °C, which is the highest value ever recorded in the 111 year observational history at that location. Furthermore, Tmax exceeding 40 °C was recorded at Hongcheon (41 °C) and Uiseong (40.4 °C), which are unprecedentedly hot temperatures in Korea. The station-averaged number of heatwaves (defined as Tmax more than 33 °C over 2 d based on the criterion for heatwave advisory issued by the KMA) in Korea in 2018 was 31.4 d (9.8 d of average per year), which is the highest since observations commenced. Not only the maximum temperature, but also the minimum temperature set a new record, with an exceptionally long period of tropical nights (i.e. 17.7 d) compared to the climatological average (i.e. 5.1 d). Although in-depth analysis and further study are required for robust attribution of the observed record-breaking extreme temperatures in 2018, it is probably inexplicable in the absence of any anthropogenic influence. Indeed, Min et al. (2014) and Kim et al. (2018) demonstrated the anthropogenic influences on the summer 2013 Korean heatwave using the fraction of attributable risk approach, in which the probability of extreme events occurring is compared in two hypothetical worlds, one without and one with human influences.

Consistent with the projected changes at the global scale such as a virtually certain increase in the frequency and magnitude of unusually warm days and nights (IPCC 2013), recent studies based on climate modeling have demonstrated that the statistical likelihood of extreme hot temperatures and heatwaves in Korea is also projected to increase due to global warming (Lee 2017, Im et al. 2015, 2017a, Lee and Min 2018, Shin et al. 2018). Most of these studies have focused on comparative analysis to show how much future climate in response to enhanced greenhouse gas emissions (GHGs) will depart from the reference climate in response to historical GHGs. However, the changing magnitude of temperature in the far future appears to be abstract and ambiguous for non-climate experts. The historical events, which had left drastic impacts, but will be a close analog to future extremes, may help improve the public recognition of the risks in a changing climate (Fitzpatrick and Dunn 2019). In this regard, characterizing the 2003 European heatwaves and identifying their statistical analog from future projections have successfully demonstrated the perception and potential threats of 'a shape of things to come' under anthropogenic warming (Beniston and Diaz 2004, Schar et al. 2004, Stott et al. 2004, Russo et al. 2014, Bador et al. 2017). For the Korean region, the 2018 summer extreme temperature is an excellent example that enables an estimation to be made of the severity of extreme events expected from different levels of global warming. In this study, we investigate the characteristics of the 2018 Korean extreme hot temperatures and compare their departure from the observational climatological normal with projected changes under 2 °C and 3 °C warmer climate conditions based on the state-of-the-art climate simulations that were generated using general circulation models (GCMs), regional climate models (RCMs) and statistical bias correction (quantile mapping). This analysis promises to explain how the unprecedentedly extreme temperatures that appeared in the observational record at the local scale can be interpreted in the context of the 2 °C and 3 °C global warming. Since the 2015 Paris Agreement, a significant body of research has quantified the impact of different levels of warming specified by the various thresholds of global average temperature (e.g. 1.5 °C, 2 °C, 3 °C) (King and Karoly 2017, Wang et al. 2017, Jacob et al. 2018, Lee and Min 2018, Lee et al. 2018, Li et al. 2018, Sylla et al. 2018). However, most of these studies applied the GCM projections with relatively coarse resolution, which is not appropriate for investigating detailed features at the regional to local scales. Studies based on dynamically downscaled projections using RCMs with higher resolution were mostly focused on the European region or a few other regions. To the best of our knowledge, Lee and Min (2018) is the only attempt to quantify the future changes in heat stress over Korea, which is centered on their study region of East Asia, under 1.5 °C and 2 °C target temperature conditions. Their study clearly revealed the mitigation benefit of half a degree less warming in terms of heat stress measured by the wet-bulb globe temperature (WBGT). However, a GCM-based result could not provide the details of localized pattern over Korea where the region-specific climate is largely influenced by complicated geographical features. For example, the Taebaek Mountains extending from north to south along the east coast of Korea distinctly differentiates the climate characteristics of the inland mountainous region from the eastern coastal region, despite their separation of less than 50 km. The GCM projections fail to capture such a spatial variability of climate variables over short distances (Qiu et al. 2019, Lee et al. 2019a). In particular, since the precise representation of elevation not only in mountainous regions, but also in low-lying basins, critically contributes to the accuracy of temperature simulation, RCM with higher resolution better resolves the sharp gradients of temperature variation and reduces the systematic biases (Qiu et al. 2019). This study based on dynamically downscaled projections with greater regional details provides a better picture of the local-scale severity of extreme temperatures and resultant heat stress that remains poorly understood. A particular focus is given to the comparison of statistical behaviors between the temperatures
observed in 2018 summer and those projected under a 2.0 °C and 3 °C warmer world. This comparison will have important implications for conceptualizing unexperienced future phenomenon and will help build upon previous findings. This is a timely study that identifies regionally emerging challenges faced by globally targeted warming and that will support the development of better adaptation strategies.

2. Experimental design and analysis method

For fine-scale climate information focusing on the Korean peninsula, three GCM projections were dynamically downscaled using three RCMs (see table 1). The Weather Research and Forecasting (WRF) and Seoul National University Mesoscale Model version 5 (SNU MM5, Cha and Lee 2009) with 12.5 km horizontal resolution were used for the dynamical downscaling of the Hadley Centre Global Environmental Model version 2—Atmosphere and Ocean (HadGEM2-AO). Baek et al (2013) demonstrated that HadGEM2-AO showed reasonable performance over East Asia, including our target domain. These two downscaled results are part of the national downsampling project in South Korea, and their detailed configuration and basic performance appear in Im et al (2015, 2017a). On the other hand, the Regional Climate Model version 4 (RegCM4, Giorgi et al 2012) with 20 km horizontal resolution was used for the dynamical downscaling of the Max Planck Institute for Meteorology Earth System Model-Mixed Resolution (MPI-ESM-MR, Giorgetta et al 2013) and Norwegian Earth System Model (NorESM1-M, Bentsen et al 2013). These two global projections are guided to give the highest priority of downscaling by the CORDEX Coordinated Output for Regional Evaluations (CORE) program (http://cordex.org/experiment-guidelines/cordex-core/). These global projections from HadGEM2-AO, MPI-ESM-MR, and NorESM1-M were forced by the representative concentration pathway (RCP8.5) that is equivalent to the business-as-usual emission scenario within the framework of the Coupled Model Intercomparison Project phase 5 (CMIP5; Taylor et al 2012).

Since the individual models have their own warming phase in response to emissions forcing, we determine the reference and two future periods that correspond to 0.48 °C, 2 °C, and 3 °C warming over the pre-industrial period (1861–1890), based on the same method used in Sylla et al (2018). Long-term observed temperature (HadCRUT.4.6, Morice et al 2012) undertaken a 30-year running mean reveals that global average temperature increased by 0.48 °C during 1976–2005 compared to that averaged over the pre-industrial period (Sylla et al 2018). The pre-industrial and reference periods were set as 1861–1890 and 1976–2005, respectively, in line with IPCC (2013). Table 1 provides the central years and averaged periods corresponding to 0.48 °C, 2 °C, and 3 °C global warming with respect to individual projections. After finding the year surpassing 0.48 °C, 2 °C, and 3 °C from the 30-year running mean of global average temperature simulated by each projection, that year is assigned as the central year for the 30-year average. This method has the advantage of considering the different sensitivity of GCMs to emission forcing, compared to applying a fixed period (e.g. 1976–2005). For reference simulation, the three projections do not cover the full 30-year period corresponding to 0.48 °C warming. For example, the downscaled results of the HadGEM2-AO projection are available from 1981; therefore only a 26 year simulation is used for representing the reference period. However, since the missing year is a marginal portion, it may not introduce a significant obstacle to establishing the validity of the conclusions presented in this study.

To validate the simulated temperature and calculate the bias correction factor using quantile mapping, we use the 56 in situ observational data maintained by the KMA during the 30-year period of 1976–2005 over South Korea (figure 1(a)). To facilitate the comparison with in situ observations, an inverse distance weighting (IDW) interpolation method is applied to convert the grid-point data of downscaled simulation into 56

### Table 1. Central years and averaged periods corresponding to the 0.48 °C, 2 °C, and 3 °C global warming derived from GCM projections forced by RCP8.5 scenario. The years marked by an asterisk indicate the starting or ending years that do not cover the full 30-year period.

<table>
<thead>
<tr>
<th>Target warming</th>
<th>RCM</th>
<th>RegCM4</th>
<th>RegCM4</th>
<th>WRF</th>
<th>SNU-MM5</th>
</tr>
</thead>
<tbody>
<tr>
<td>Reference (0.48 °C)</td>
<td>Central year</td>
<td>1978</td>
<td>1991</td>
<td>1991</td>
<td>1991</td>
</tr>
</tbody>
</table>

| Future (2 °C) | Central year | 2038   | 2048   | 2042    | 2042    |
|               | Period       | 2024–2053 | 2034–2063 | 2028–2057 | 2028–2057 |
|               | Length       | 30-year  | 30-year | 30-year  | 30-year  |
| Future (3 °C) | Central year | 2060   | 2073   | 2061    | 2061    |
|               | Period       | 2046–2075 | 2059–2088 | 2047–2076 | 2047–2076 |
|               | Length       | 30-year  | 30-year | 30-year  | 30-year  |
stations. The closest four points from the corresponding station are used for the IDW interpolation. In particular, we focus on three stations: Seoul, Uiseong, and Daegu (figure 1(b)). While Seoul and Uiseong experienced extraordinary hot temperatures in summer 2018, Daegu is historically recognized as one of the hottest cities in Korea due to its geographical setting characterized by a low-lying basin surrounded with mountains (Im et al 2017a).

Although dynamical downsampling using RCMs can provide added value over driving GCMs in terms of resolving the heterogeneity of complex geographical features, any climate simulations include unavoidable biases, which reduce the reliability of quantitative estimates in future changes. For example, the extreme analysis counting the exceedance above the absolute threshold may be affected by systematic cold or warm biases. Therefore, statistical bias correction using quantile mapping is applied to remove the systematic bias embedded in the downscaled raw results. The quantile mapping matches the cumulative distribution function (CDF) of raw outputs to observed CDF,
which allows the whole distribution to be effectively adjusted, thereby improving the mean, variance, and extreme values (Lee et al. 2019b). The Gaussian distribution is used to estimate the probability distribution function for daily Tmax with respect to 56 observational in situ stations. A goodness of fit for the estimated distribution for individual stations is verified by the Kolmogorov–Smirnov (K–S) test (Conover 1999), and they are statistically significant at the 95% confidence level for most of stations, which supports the suitability of using the Gaussian distribution for fitting the simulated distributions. To evaluate the bias-corrected reference simulation, the frequency distribution of daily Tmax derived from individual four projections after bias correction and their ensemble means are compared at the station level against in situ observation data (figure S1 is available online at stacks.iop.org/ERL/14/094020/mmedia). In addition to the distribution of all 56 stations, three stations (Seoul, Daegu and Uiseong) are individually presented because they are a particular focus in the following analysis. Although individual projections show somewhat different performance, they all are in good agreement with the observed pattern, with the capability to reproduce the present climate. For more quantitative assessments, we provide the metric of the two-sample K–S distance that indicates the maximum distance between two (simulated versus observed) CDFs. Figure S2 presents the K–S distance between simulated and observed daily Tmax distribution for July and August at 56 stations. Daily Tmax distributions for the majority of stations show that their K–S distances do not exceed the critical value at the 99% level, which indicates a certain level of similarity between the simulated and observed distributions. The reasonable performance in capturing the climatological characteristics for the reference period is crucial to enhance the reliability of future climate projections. Under the assumption of ‘stationarity’, which means that the bias pattern does not change with time, the bias correction factor computed from the reference simulation is equally applied to the future projections.

Among the various heat stress indices that measure the combined effects of temperature and humidity, WBGT is used in this study. As presented by table 2 in Willett and Sherwood (2012), WBGT provides a clear threshold to describe the level of risk to physical activity. Due to the unavailability of black globe temperature, we calculate the simplified WBGT, which is an approximation to the WBGT that assumes moderately high radiation levels and light wind conditions. Therefore, the formula to calculate the simplified WBGT includes only temperature and humidity, without accounting for the effect of radiative fluxes and wind (Willett and Sherwood 2012), which is widely accepted (hereafter referred to as WBGT).

\[
WBGT = 0.567T_a + 0.393e + 3.94
\]

\[
e = (RH/100) \times 6.105 \exp \left( \frac{17.27T_a}{237.7 + T_a} \right),
\]

where \(T_a\) is the surface air temperature (°C), \(e\) is the vapor pressure (hPa), and RH is the relative humidity (%).

3. Results

We begin our analysis with the spatial distribution of daily Tmax averaged over July and August derived from observational station data: 30-year climatological mean (1976–2005) versus 2018 single year (figures 1(a), (b)). The 56 station data are interpolated into a grid using an IDW interpolation method because the spatial pattern is easily visible from the gridded distribution, but we also provide the same quantity of in situ observational sites without any interpolation (figure S3). Comparing these two patterns gives insight into the extent to which 2018 summer deviated from the normal condition. To clearly demonstrate the regional departure pattern from climatology, figure 1 also presents the spatial distribution of the anomaly calculated by subtracting the climatology from 2018 Tmax. In summer 2018, a Tmax anomaly above at least 2 °C appears over the entire region of South Korea (figure 1(c)). Such a magnitude of anomalously high temperature is roughly comparable with the plus two standard deviation range of Tmax (figure 1(d)), although the regional pattern shows little consistency between the anomaly and standard deviation. In particular, two localized maxima covering the Seoul and Uiseong stations reach a magnitude of more than 3.5 °C (figure 1(c)). To identify the severity of this anomalous pattern, we compare the 2018 Tmax anomaly with the future changes in Tmax derived from the ensemble mean of four downscaled projections (see table 1, hereafter referred to as ENS). Future changes are calculated by the difference between 2 °C (3 °C) warming and reference periods as denoted in table 1 with respect to four projections individually, and they are then averaged as ENS. The comparison between the 2018 observed anomaly and Tmax increase from future projections demonstrates that the anomalous warming appearing in some regional maxima of summer 2018 already overwhelms the magnitude of the Tmax increase expected under 2 °C global warming. Although the 3 °C global warming scenario produces a much larger increase of Tmax along the eastern and southern coastal regions, some regions seem to be comparable with those in the 2018 anomalous magnitude. While Tmax averaged over South Korea is projected to increase by 2.7 °C under the 2 °C target of global average temperature increase, this increment is greatly increased to 4.2 °C under 3 °C global warming. This implies that regional warming
may not necessarily be estimated by linear scaling from global warming. On the gross pattern of Tmax averaged over the summer season (e.g. July and August), we examine the frequency distribution of daily Tmax at Seoul, Daegu and Uiseong individual stations, as well as at all 56 stations (figure 2). For this analysis, we pool together all Tmax events occurring during July and August for each bin with empirically chosen intervals of 0.5 °C. The frequencies counted for the 30-year period (i.e. OBS Clim) are divided by the number of years (i.e. 30) in order to directly compare the results from a single year (i.e. 2018). For the RCM simulations, the results from four projections are pooled together without averaging. The vertical lines indicate the average value corresponding to the distribution with the same color (solid lines for RCM, but dashed lines for OBS). The curves denoted by 'OBS Clim' and 'RCM Ref' in this figure are the same as 'OBS Clim' and 'All RCMs' in figure S1, respectively.

Figure 2. Frequency distribution of daily Tmax (°C) during July and August at Seoul (a), Daegu (b), Uiseong (c), and all 56 stations (d). All Tmax values occurring in the target period are pooled and then divided by the number of years with respect to each bin. The results from four projections are pooled together without averaging. The vertical lines indicate the average value corresponding to the distribution with the same color (solid lines for RCM, but dashed lines for OBS). The curves denoted by 'OBS Clim' and 'RCM Ref' in this figure are the same as 'OBS Clim' and 'All RCMs' in figure S1, respectively.
shape at particular stations (figures 2(a)–(c)). The distribution from all stations shows stronger similarity between the distribution derived from 2018 Tmax and that expected from 3 °C global warming. This supports that the uncommonly high Tmax under the current climate could become characterized as the new normal in the future if the global average temperature is allowed to increase by up to 3 °C. Similar findings have been reported, dealing with climate change issues in the European region (Beniston 2004, Beniston and Diaz 2004, Schär et al 2004, Stott et al 2004, Russo et al 2014).

The unprecedented severity seen in the 2018 extreme temperatures is further explored using the CDF, which readily shows the probability of Tmax exceeding certain thresholds. Figure 3 presents the CDF of Tmax at Seoul, Daegu and Uiseong stations, as well as at all 56 stations derived from observations and RCM simulations. By comparison with the observed climatological pattern, the 2018 CDF exhibits a systematic shift to higher temperatures across all percentiles. The statistical behaviors of Tmax at Seoul and Uiseong stations are particularly relevant. The median CDF values (i.e. the 50th percentile) reach 33.8 °C and 35.7 °C at Seoul and Uiseong in summer 2018, respectively. These are larger than the arithmetic mean values presented in figure 2, and describe a distribution with a negatively skewed shape. The significant changes in median value well explain the extraordinary nature of the 2018 hot temperatures. More specifically, Tmax at Seoul (Uiseong) during 1976–2005 does not exceed 33.8 °C (35.7 °C), which is the 50th percentile in summer 2018, until the cumulative probability reaches the 95th (97th) percentile. In other words, Tmax occupying the very upper rank (i.e. 5th or 3rd percentile) in the normal condition becomes the central tendency (i.e. 50th percentile) in summer 2018. Consequently, the intensity of extreme heat events, which are typically measured by Tmax exceeding the 95th percentile, also dramatically increases to 38.0 °C and 39.8 °C at Seoul and Uiseong, respectively. In addition, Tmax ranging approximately from the 30th to 95th percentiles largely surpasses that from 3 °C global warming at Seoul and Uiseong, indicating the substantial severity of Tmax in 2018. Moving to the comparison of RCM simulations generated with different levels of increase in global average temperature (i.e. 0.48 °C, 2 °C and 3 °C), global warming manifests its impacts through shifting the distributions towards higher temperatures compared to the reference climate, which is commonly found in future projections (e.g. Im et al 2015). The CDFs under 3 °C global warming at three stations mostly resemble that from summer 2018. Consistent with the frequency distribution in figure 2, this analysis further suggests that the significantly distinct behavior of the 2018 extreme temperatures will probably become the norm in Korea under 3 °C global warming.
An increase in temperatures will lead to more frequent and more extended heatwaves. The persistence of extreme hot temperatures can further exacerbate the negative impacts in many sectors (e.g. public health) due to the accumulation of the impacts. Figure 4 presents the frequency distribution of consecutive hot days with Tmax exceeding 35 °C at various durations at 56 stations. The KMA has issued a heatwaves advisory notice and a warning when the daily Tmax is expected to exceed 33 °C and 35 °C for at least 2 d, respectively. Based on this criterion of heatwaves warning, heatwaves with different durations are counted using 62 d (July-August) Tmax for every year at 56 stations. The 2018 observation is the result from only one-year, but others are normalized by dividing the total number of years with respect to each duration. For the RCM simulation, the color bars indicate the four ensemble means while the upper and lower bars indicate the maximum and minimum values of the corresponding ensemble members, respectively.

Figure 4. Frequency distribution of heatwaves at various durations derived from 30-year (1976–2005) and one-year (2018) observations and four projections under 0.48 °C (i.e. Reference), 2 °C and 3 °C global warming. Except for the single-year result (i.e. 2018), all others are normalized by dividing the total number of years with respect to each duration. For the RCM simulation, the color bars indicate the four ensemble means while the upper and lower bars indicate the maximum and minimum values of the corresponding ensemble members, respectively.

Among the detrimental consequences of more frequent and intensified extreme hot temperatures, human mortality and morbidity are probably the greatest concern (Son et al 2012, Mitchell et al 2016). Indeed, the South Korean Ministry of Health, Welfare and Disease Control announced that at least 29 people died due to heatstroke alone in summer 2018, supporting the acutely adverse impact of extreme hot temperatures on human health. From the perspective of health risk, temperature is not the only factor that can affect thermal discomfort and humans’ ability to control body temperature. It is also equally important to consider ambient humidity because humans are sensitive to the combination of temperature and humidity (Steinweg and Gutowski 2015, Pal and Eltahir 2016, Im et al 2017b). Therefore, we also investigate the characteristics of WBGT that is widely used to measure the heat stress and accompanied risk level (Willett and Sherwood 2012, Im et al 2017a, Lee and Min 2018). Figure 5 presents WBGT and daily mean temperature and daily mean relative humidity (RH) used to calculate WBGT. Individual points indicate
the median value obtained from the distribution at each station. For each station, the distribution of 2018 observation is made up of 62 d values while the distribution of climatological observation is made up of $62 \times 30$-year values. WBGT is not computed using daily Tmax because the KMA does not provide the RH data measured when daily Tmax occurs. Since the diurnal cycle of RH is roughly out of phase with temperature, the WBGT calculation using temporally unmatched RH and Tmax might induce a nontrivial error. Therefore, WBGT is calculated using daily mean temperature and daily mean RH, which leads to an underestimation of the heat stress severity compared to WBGT calculated using Tmax and simultaneous RH. Nevertheless, this analysis may well illustrate how the changes in temperature influence the risk arising from heat stress. In general, WBGT and temperature are positively correlated. Therefore, the higher WBGT takes place along the higher temperature accordingly. In Korea’s typical summer climate, WBGT mostly ranges from 26°C to 29°C, which corresponds to a moderate or marginally high heat stress risk. For summer 2018, the risk level sharply increases and enters the extreme risk category. Although some stations show reduced RH, WBGT is still greatly amplified because the increase in temperature is large enough to offset the effect of reduced RH. Under 3°C global warming, many more stations are likely to experience extreme risk level, even though some stations are positioned within a similar range with that obtained from the 2018 meteorological condition. The statistics (e.g. median) derived from single-year (62 d) and 30-year ($62 \times 30$ years $\times 4$ members) periods may not agree perfectly. Although the median is considered to represent the typical value of the distribution, the inter-annual variability during the 30-year period may influence the result. In spite of this caveat, comparing WBGT calculated from the 2018 observational data with RCM projections under 3°C global warming may help to figure out the specific consequences that a particular region will face due to a certain level of global warming. More importantly, the emergence of far more extreme risks that have never previously arisen in even record-breaking extreme temperatures (e.g. 2018 summer) enables an estimation to be made of the severity of extreme events expected from 3°C global warming. Most people in Korea will be exposed to severely hazardous summers in terms of heat stress if we fail to limit the global average temperature increase to well below 2°C above the pre-industrial levels.

4. Discussion and conclusions

It is largely considered that the risks of extreme maximum temperatures and resultant heat stress will continue to increase over many different regions in tandem with the rise in global average temperature (Bador et al 2017, Im et al 2017a, 2017b, Lee and Min 2018, Sylla et al 2018). However, the severity of extreme temperature events at the regional to local scale may not be necessarily proportional to the degree of global warming (Knutti et al 2016, Harrington et al 2018). Their temporal and spatial patterns still remain uncertain. Since climate change is a global issue with region-specific impacts, it is imperative to quantify the
changes in local temperature in response to the threshold of the globally aggregated target (e.g. 2°C, 3°C) in order to better understand the regional consequences of either limiting or allowing further global warming.

In this study, the extraordinary nature of the 2018 extreme hot temperatures is characterized against the climatological observational pattern, and compared with the dynamically downscaled fine-scale climate projections under 2°C and 3°C global warming scenarios above the pre-industrial level. The in situ observations in July and August 2018 clearly demonstrate substantially higher Tmax across all of South Korea, and the anomalous warmings that appeared in some regions overwhelm the plus two standard deviation range of historical variation of Tmax averaged over July and August. Accordingly, the frequency distribution of Tmax in 2018 departs significantly from the climatological pattern. Both the mean Tmax and the distribution’s shape differ substantially from the typical Gaussian fit that is normally applied to describe the behavior of temperature. In particular, the asymmetric distribution with very heavy and skewed upper tail well explains the increase in probability of extreme Tmax. For example, Tmax at Seoul during 1976–2005 does not exceed 35°C until the cumulative probability reaches about the 98th percentile, whereas Tmax of 35°C is only positioned at the 64th percentile in summer 2018 because the much higher Tmax is placed in the upper rank. The increase in Tmax directly results in a substantial increase of the frequency and durations of heatwaves. The strong positive relationship observed between Tmax and WBGT indicates that a significant increase in Tmax can induce a high risk of heat stress even when RH is reduced, which is in line with the findings of Im et al. (2018).

Comparing the 2018 anomalous pattern and future climate projections generated under different levels of global warming gives us a better idea of how to interpret the severity of the 2018 summer extreme temperatures within the context of 2°C and 3°C global warming. Locally tailored climate projections obtained through dynamical downscaling and statistical bias correction can capture the geographically diverse characteristic of Tmax over Korea, showing a good agreement with the historical observed pattern. Statistically, the 2018 Tmax and resultant heat stress closely resemble the climate projections expected under the 3°C global warming scenario. The statistical analog of Tmax between 2018 observations and RCM projections under 3°C global warming suggests that the unprecedented severity of extreme heat occurring in the current climate could become the future norm unless the increase in global average temperature is stabilized at an acceptable level. A similar conclusion has been drawn from a comparison of the 2003 European heatwaves and future projection in response to elevated GHG forcings (e.g. Beniston 2004, Beniston and Díaz 2004, Schär et al 2004, Stott et al 2004, Russo et al 2014, Bador et al 2017).

While characterizing the historical extreme events and assessing the statistical likelihood of future extreme events give insight into the severe impacts of climate change, further in-depth studies are needed to extend our understanding of the physical mechanisms contributing to the changes in intensity and frequency of extreme events. The possible mechanisms behind extreme heatwaves in Korea are complex and have not been fully explored (Kim et al 2019). However, the extent and strength of the western North Pacific subtropical high around Korea are often investigated to explain the physical mechanism that is responsible for the heatwaves in South Korea (Yeh et al 2018, Yoon et al 2018). This is because extreme heatwaves are linked to a specific atmospheric circulation at the synoptic scale (Lee and Lee 2016), even though they are a locally dominated phenomenon. Figure S6 presents the spatial distribution of the historical climatology of the geopotential height at 500 hPa and its anomalous pattern dominated by the year with extreme heatwaves during July and August. The three years with extreme heatwaves under 3°C global warming that seem to be equivalent to the 2018 case are selected from the temporal evolution of the Tmax anomaly (figure S5). This analysis clearly demonstrates the synoptic framing of the atmospheric condition favorable for heatwaves and the potential of GCM projections to capture these behaviors. In summer 2018, a strong positive anomalous pattern of the geopotential height is found because the significant expansion of the western North Pacific subtropical high to the westward provokes an anomalous anticyclonic circulation over Korea (figure S6(a)). Similarly, the synoptic condition for extreme heatwaves projected under 3°C global warming is characterized by the enhanced high-pressure system (figure S6(b)), which infers a persistent large-scale subsidence that is favorable for inducing prolonged and intense heatwaves (Lee and Lee 2016). Since the tropospheric warming leads to the greater geopotential height due to the thermal expansion of the atmosphere, the climatology of geopotential height under 3°C global warming is much higher than that of historical period. Therefore, the anomaly of GCM ENS against historical climatology shows much higher values, compared to observed anomaly of 2018. On the other hand, the magnitude of anomaly against 3°C warmer climatology is smaller than that of 2018 because the selected extremes are no longer extraordinary ones under 3°C warmer climate. However, the anomalous spatial pattern of geopotential height expected from extreme heatwaves shows a large similarity with observed one despite some discrepancy in the spatial details (e.g. the position of the maximum). This level of similarity between GCM ENS and ERA-Interim reanalysis is sufficient to support the reliability of GCMs that provide the proper large-scale forcing to the lateral boundary while running RCM. In addition, further research should focus on attributing temperature extremes to...
anthropogenic climate change, which will enhance the confidence level of their changed probability under global warming. Very few studies have so far attempted to detect an anthropogenic influence in extreme heatwaves over South Korea or to analyze their attribution to either external forcing or internal climate variability (e.g. Min et al. 2014).

Although this study has focused on the extreme hot temperatures occurring in Korea, much of our planet suffered a sizzling summer in 2018 with record-breaking temperatures worldwide. The unprecedented 2018 extreme temperatures appears to have enhanced the public awareness of the potentially drastic effects of climate change, beyond being an abstract concept. It should be considered a timely warning of the need for global action to successfully achieve the Paris Agreement on climate change.

Acknowledgments

Im E-S and Nguyen-Xuan T were supported by the Korea Environment Industry & Technology Institute (KEITI) through the Advanced Water Management Research Program, which is funded by the Korea Ministry of Environment (MOE) (Grant 83079). Ahn J-B and Kim Y-H were supported by the Korea Meteorological Administration Research and Development Program under Grant KMI2018-01213.

Data availability statement

The data that support the findings of this study are available from the corresponding author upon reasonable request.

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