



1 **Trends in evaporative demand in Great Britain using high-**  
2 **resolution meteorological data**

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4 **Emma L. Robinson<sup>1</sup>, Eleanor M. Blyth<sup>1</sup>, Douglas B. Clark<sup>1</sup>, Jon Finch<sup>1</sup> and Alison**  
5 **C. Rudd<sup>1</sup>**

6 [1]{Centre for Ecology and Hydrology, Maclean Building, Benson Lane, Crowmarsh Gifford,  
7 Wallingford OX10 8BB}

8 Correspondence to: Emma L. Robinson (emrobi@ceh.ac.uk)

9



1 **Abstract**

2 Observations of climate are often available on very different spatial scales from observations  
3 of the natural environments and resources that are affected by climate change. In order to help  
4 bridge the gap between these scales using modelling, a new dataset of daily meteorological  
5 variables was created at 1 km resolution over Great Britain for the years 1961-2012, by  
6 interpolating coarser resolution climate data and including the effect of local topography. These  
7 variables were used to calculate evaporative demand at the same spatial and temporal  
8 resolution, both excluding (PET) and including (PETI) the effect of water intercepted by the  
9 canopy. Temporal trends in evaporative demand were calculated, with PET found to increase  
10 in all regions and PETI found to increase in England. The trends were found to vary by season,  
11 with spring evaporative demand increasing by 14% (11% when the interception correction is  
12 included) in Great Britain over the dataset, while there is no statistically significant trend in  
13 other seasons. The trends in PET were attributed analytically to trends in the climate variables,  
14 with the spring trend in evaporative demand being driven by radiation trends, particularly by  
15 increasing solar radiation.

16



## 1 1 Introduction

2 There are many studies showing the ways in which our living environment is changing over  
3 time: wildlife surveys in the UK of both flora (Wood et al., 2015; Evans et al., 2008) and fauna  
4 (Pocock et al., 2015) show a shift in patterns and timing (Thackeray et al., 2010). In addition,  
5 the UK natural resources of freshwater (Watts et al., 2015), soils (Reynolds et al., 2013;  
6 Bellamy et al., 2005) and vegetation (Berry et al., 2002; Hickling et al., 2006; Norton et al.,  
7 2012) are changing. We are experiencing new environmental stresses on the land and water  
8 systems of the UK through changes in temperature and rainfall (Crooks and Kay, 2015; Watts  
9 et al., 2015; Hannaford, 2015).

10 To explain these changes in terms of physical drivers, there are several gridded meteorological  
11 datasets available for Great Britain. Some are derived directly from observations – for example  
12 the Met Office Rainfall and Evaporation Calculation System (MORECS) dataset (Thompson et  
13 al., 1981; Hough and Jones, 1997), the UKCP09 observed climate data (Jenkins et al., 2008)  
14 and the Climate Research Unit time series 3.21 (CRU TS 3.21) data (Jones and Harris, 2013;  
15 Harris et al., 2014) – while some use global meteorological reanalyses bias-corrected to  
16 observations – for example the WATCH Forcing Data (WFD; Weedon et al. (2011)), the  
17 WATCH Forcing Data methodology applied to ERA-Interim reanalysis product (WFDEI;  
18 Weedon et al. (2014)) and the Princeton Global Meteorological Forcing Dataset (Sheffield et  
19 al., 2006).

20 However, while observations of carbon, methane and water emissions from the land (Baldocchi  
21 et al., 1996), the vegetation cover (Morton et al., 2011) and soil properties  
22 (FAO/IIASA/ISRIC/ISS-CAS/JRC, 2012) are typically made at the finer landscape scale of  
23 100 m to 1000 m, most long-term meteorological datasets are only available at a relatively  
24 coarse resolution of a few tens of km. These spatial scales may not be representative of the  
25 climate experienced by the flora and fauna being studied, and it has also been shown that input  
26 resolution can have a strong effect on the performance of hydrological models (Kay et al.,  
27 2015). In addition, the coarse temporal resolution of some datasets, for example the monthly  
28 CRU TS 3.21 data (Harris et al., 2014; Jones and Harris, 2013), can miss important sub-monthly  
29 extremes. It is imperative for our increased understanding and improved analysis of the  
30 environment that we bridge the gap between the scales of observations with modelling.  
31 However, while there are datasets available at higher spatial and temporal resolutions (such as



1 UKCP09 (Jenkins et al., 2008)), these often do not provide all the variables needed for land  
2 surface or hydrological modelling.

3 To address this, we have created a meteorological dataset for Great Britain at 1 km resolution  
4 (Robinson et al., 2015b). It is derived from the observation-based MORECS dataset (Thompson  
5 et al., 1981; Hough and Jones, 1997), which is downscaled using information about topography.  
6 This is augmented by an independent precipitation dataset – Gridded Estimates of daily and  
7 monthly Areal Rainfall for the United Kingdom (CEH-GEAR; Tanguy et al. (2014); Keller et  
8 al. (2015)) – along with variables from two global datasets – WFD Weedon et al. (2011) and  
9 CRU TS 3.21 (Harris et al., 2014; Jones and Harris, 2013) – to produce a comprehensive,  
10 observation-based, daily meteorological dataset at 1 km × 1 km spatial resolution.

11 In addition, a key variable in hydrological modelling is the evaporative demand of the  
12 atmosphere, which is determined by the meteorological variables (Kay et al., 2013).  
13 Hydrological models such as Climate and Land use Scenario Simulation in Catchments  
14 (CLASSIC; Crooks and Naden (2007)) or Grid-to-Grid (G2G; Bell et al. (2009)), and metrics  
15 such as the Palmer Drought Severity Index (PDSI; Palmer (1965)) require potential  
16 evapotranspiration (PET) – an estimate of the unstressed evaporative demand of the atmosphere  
17 – as an input. While hydrological models can make use of high resolution topographic  
18 information and precipitation datasets, they are often driven with PET calculated at a coarser  
19 resolution (Bell et al., 2011; Bell et al., 2012; Kay et al., 2015). Therefore, we have also created  
20 a dataset consisting of two estimates of PET, which can be used to run high-resolution  
21 hydrological models (Robinson et al., 2015a).

22 This paper presents the method of creation of the new high-resolution meteorological and PET  
23 datasets. Regional trends in evaporative demand are then calculated and attributed to regional  
24 trends in the meteorological data.

## 25 **2 Calculation of meteorological variables**

26 The meteorological variables included in this new dataset (Robinson et al., 2015b) are air  
27 temperature, specific humidity, wind speed, downward long- and shortwave radiation,  
28 precipitation, daily temperature range and air pressure (Table 1). These variables are important  
29 drivers of near-surface conditions, and are the full set of variables required to drive the JULES  
30 land surface model (LSM) (Best et al., 2011; Clark et al., 2011).



1 The data were derived primarily from MORECS, which is a long-term gridded dataset starting  
2 in 1961 and updated to the present (Thompson et al., 1981; Hough and Jones, 1997). It  
3 interpolates five daily synoptic station variables (air temperature, vapour pressure, wind speed,  
4 hours of bright sunshine, rainfall) to a 40 km × 40 km resolution grid aligned with the Ordnance  
5 Survey National Grid. The interpolation is such that the value in each grid square is the effective  
6 measurement of a station positioned at the centre of the square and at the grid square mean  
7 elevation, averaged from 09:00 GMT to 09:00 GMT the next day. MORECS is a consistent,  
8 quality-controlled time series, which accounts for changing station coverage. The MORECS  
9 variables were used to derive the air temperature, specific humidity, wind speed, downward  
10 long- and shortwave radiation and air pressure in the new dataset. The WFD and CRU TS 3.21  
11 datasets were used where variables could not be calculated solely from MORECS, except for  
12 precipitation, for which the CEH-GEAR data were used instead of interpolating the MORECS  
13 rainfall.

14 The spatial coverage of the dataset was determined by the spatial coverage of MORECS, which  
15 covers the majority of Great Britain, but excludes some coastal regions and islands at the 1 km  
16 scale. For many of these points, the interpolation was extended from the nearest MORECS  
17 squares, but some outlying islands (in particular Shetland and the Scilly Isles) were deemed to  
18 be too far from any MORECS squares and were therefore excluded.

## 19 **2.1 Air temperature**

20 Air temperature,  $T_a$  (K), was derived from the MORECS air temperature. The MORECS air  
21 temperature was reduced to mean sea level, using a lapse rate of  $-0.006 \text{ K m}^{-1}$  (Hough and  
22 Jones, 1997). A bicubic spline was used to interpolate from 40 km resolution to 1 km resolution,  
23 then the temperatures were adjusted to the elevation of each 1 km square using the same lapse  
24 rate. The 1 km resolution elevation data were aggregated from the Integrated Hydrological  
25 Digital Terrain Model (IHDTM) – a 50 m resolution digital terrain model (Morris and Flavin,  
26 1990).

## 27 **2.2 Specific humidity**

28 Specific humidity,  $q_a$  ( $\text{kg kg}^{-1}$ ), was derived from the MORECS vapour pressure, which was  
29 first reduced to mean sea level, using a lapse rate of  $-0.025 \text{ \% m}^{-1}$  (Hough and Jones, 1997). A  
30 bicubic spline was used to interpolate to 1 km resolution then the vapour pressure values were



1 adjusted to the 1 km resolution elevation using the IHDTM elevations (Sect. 2.1). Finally the  
2 specific humidity was calculated, assuming a constant air pressure,  $p^* = 100000$  Pa, using

$$3 \quad q_a = \frac{\epsilon e}{p^* - (1 - \epsilon)e}, \quad (1)$$

4 where  $e$  is the vapour pressure (Pa) and  $\epsilon = 0.622$  is the mass ratio of water to dry air (Gill,  
5 1982).

### 6 **2.3 Downward shortwave radiation**

7 Downward shortwave radiation,  $S_d$  ( $\text{W m}^{-2}$ ), was derived from the MORECS hours of bright  
8 sunshine. The value calculated is the mean shortwave radiation over 24 hours. The sunshine  
9 hours were used to calculate the cloud cover factor,  $C_f = n/N$ , where  $n$  is the number of hours  
10 of bright sunshine in a day, and  $N$  is the total number of hours between sunrise and sunset. The  
11 cloud cover factor was interpolated to 1 km resolution using a bicubic spline. The downward  
12 shortwave solar radiation for a horizontal plane at the Earth's surface was calculated using the  
13 solar angle equations of Iqbal (1983) and a form of the Angstrom-Prescott equation which  
14 relates hours of bright sunshine to solar irradiance (Ångström, 1918; Prescott, 1940), with  
15 empirical coefficients calculated by Cowley (1978). The Cowley coefficients vary spatially and  
16 seasonally and effectively account for reduction of irradiance with increasing solar zenith angle,  
17 as well as implicitly accounting for spatially- and seasonally-varying aerosol effects. However,  
18 they do not vary interannually and thus do not explicitly include long-term trends in aerosol  
19 concentration.

20 In addition, the downward shortwave radiation was corrected for the average inclination and  
21 aspect of the surface, assuming that only the direct beam radiation is a function of the inclination  
22 and that the diffuse radiation is homogeneous. It was also assumed that the cloud cover is the  
23 dominant factor in determining the diffuse fraction (Muneer and Munawwar, 2006). The aspect  
24 and inclination were calculated using the IHDTM elevation at 50 m resolution, following the  
25 method of Horn (1981), and were then aggregated to 1 km resolution. The top of atmosphere  
26 flux for horizontal and inclined surfaces was calculated following Allen et al. (2006) and the  
27 ratio used to scale the direct beam radiation.



## 1    **2.4 Downward longwave radiation**

2    Downward longwave radiation,  $L_d$  ( $\text{W m}^{-2}$ ), was derived from the 1 km resolution air  
3    temperature (Sect. 2.1), vapour pressure (Sect. 2.2) and cloud cover factor (Sect. 2.3). The  
4    downward longwave radiation for clear sky conditions was calculated as a function of air  
5    temperature and precipitable water using the method of Dilley and O'Brien (1998), with  
6    precipitable water calculated from air temperature and humidity following Prata (1996). The  
7    additional component due to cloud cover was calculated using the equations of Kimball et al.  
8    (1982), assuming a constant cloud base height of 1000 m.

## 9    **2.5 Wind speed**

10    The 10 m wind speed,  $u_{10}$  ( $\text{m s}^{-1}$ ), was derived from the MORECS 10 m wind speed. The  
11    MORECS wind speed data were interpolated to 1 km resolution using a bicubic spline and  
12    adjusted for topography using a 1 km resolution dataset of mean wind speeds produced by the  
13    UK Energy Technology Support Unit (ETSU) (Newton and Burch, 1985; Burch and  
14    Ravenscroft, 1992). This used Numerical Objective Analysis Boundary Layer (NOABL)  
15    methodology and station wind measurements over the period 1975-84 to produce a map of  
16    mean wind speed over the UK. To calculate the topographic correction, the ETSU wind speed  
17    was aggregated to 40 km resolution, then the difference between each 1 km value and the  
18    corresponding 40 km mean found. This difference was added to the interpolated daily wind  
19    speed.

## 20    **2.6 Precipitation**

21    Precipitation,  $P$  ( $\text{kg m}^{-2} \text{s}^{-1}$ ), is taken from the daily CEH-GEAR dataset (Tanguy et al., 2014;  
22    Keller et al., 2015), scaled to the appropriate units. The CEH-GEAR methodology uses natural  
23    neighbour interpolation to interpolate synoptic station data to a 1 km resolution gridded daily  
24    precipitation dataset of the estimated rainfall in 24 hours between 09:00 GMT and 09:00 GMT  
25    the next day.

## 26    **2.7 Daily temperature range**

27    Daily temperature range (DTR),  $D_T$  (K), was obtained from the CRU TS 3.21 monthly mean  
28    daily temperature range estimates on a  $0.5^\circ$  latitude  $\times$   $0.5^\circ$  longitude grid, which is interpolated  
29    from monthly climate observations (Harris et al., 2014; Jones and Harris, 2013). These data



1 were reprojected to the 1 km grid with no interpolation and the monthly mean used to populate  
2 the daily values in each month.

### 3 **2.8 Surface air pressure**

4 Surface air pressure,  $p^*$  (Pa), was derived from the WFD, an observation-corrected reanalysis  
5 product, which provides 3 hourly meteorological data for 1958-2001 on a  $0.5^\circ$  latitude  $\times$   $0.5^\circ$   
6 longitude resolution grid (Weedon et al., 2011). Mean monthly values of WFD surface air  
7 pressure and air temperature were calculated for each  $0.5^\circ$  grid box over the years 1961-2001.  
8 These were reprojected to the 1 km grid with no interpolation, then the air temperature used to  
9 lapse the air pressure from the WFD elevation to the 1 km resolution elevation using the  
10 temperature lapse rate specified in Sect. 2.1 (Shuttleworth, 2012). The mean monthly values  
11 were used to populate the daily values in the full dataset, thus the surface air pressure in the  
12 new dataset does not vary interannually. This is reasonable as the trend in surface air pressure in  
13 the WFD is negligible.

### 14 **2.9 Spatial and seasonal patterns of meteorological variables**

15 Long-term mean values of each meteorological variable were calculated for each 1 km square  
16 over the whole dataset (1961-2012) (Fig. 1). Mean-monthly climatologies (Fig. 2) were  
17 calculated over the whole of Great Britain (GB), and over four sub-regions of interest (Fig. 3).  
18 Three of these regions correspond to the nations (England, Wales and Scotland), while the  
19 fourth is the 'English lowlands', a subset of the English region, which includes south-central  
20 and south-east England, East Anglia and the East Midlands (Folland et al., 2015).

21 The maps clearly show the effect of topography on the variables (Fig. 1), with an inverse  
22 correlation between elevation and temperature, specific humidity, downward longwave  
23 radiation and surface air pressure and a positive correlation with wind speed. The precipitation  
24 has an east-west gradient due to prevailing weather systems and orography. The fine-scale  
25 structure of the downward shortwave radiation is due to the aspect and elevation of each grid  
26 cell, with more spatial variability in areas with more varying terrain. As no topographic  
27 correction has been applied to DTR, it varies only on a larger spatial scale.

28 The mean-monthly climatologies (Fig. 2) demonstrate the differences between the regions, with  
29 Scotland generally having lower temperatures and more precipitation than the average, and  
30 England (particularly the English lowlands) being warmer and drier.





### 1 3 Calculation of potential evapotranspiration (PET)

2 The Penman-Monteith PET,  $E_P$  (mm d<sup>-1</sup>, equivalent to kg m<sup>-2</sup> d<sup>-1</sup>), is a physically-based  
3 formulation of the evaporative demand of the atmosphere (Monteith, 1965). It is calculated  
4 from the daily meteorological variables using the equation

$$5 \quad E_P = \frac{t_d}{\lambda} \frac{\Delta A + \frac{c_p \rho_a}{r_a} (q_s - q_a)}{\Delta + \gamma \left(1 - \frac{r_s}{r_a}\right)}, \quad (2)$$

6 where  $t_d = 86400$  s d<sup>-1</sup> is the length of a day,  $\lambda = 2.5 \times 10^6$  J kg<sup>-1</sup> is the latent heat of evaporation,  
7  $q_s$  is saturated specific humidity (kg kg<sup>-1</sup>),  $\Delta$  is the gradient of saturated specific humidity with  
8 respect to temperature (kg kg<sup>-1</sup> K<sup>-1</sup>),  $A$  is the available energy (W m<sup>-2</sup>),  $c_p = 1010$  J kg<sup>-1</sup> K<sup>-1</sup> is the  
9 specific heat capacity of air,  $\rho_a$  is the density of air (kg m<sup>-3</sup>),  $q$  is specific humidity (kg kg<sup>-1</sup>),  
10  $\gamma = 0.004$  K<sup>-1</sup> is the psychrometric constant,  $r_s$  is stomatal resistance (s m<sup>-1</sup>) and  $r_a$  is aerodynamic  
11 resistance (s m<sup>-1</sup>) (Stewart, 1989).

12 The saturated specific humidity,  $q_s$  (kg kg<sup>-1</sup>), is calculated from saturated vapour pressure,  $e_s$   
13 (Pa), using Eq. (1). The saturated vapour pressure is calculated using an empirical fit to air  
14 temperature

$$15 \quad e_s = p_s \exp\left(\sum_{i=1}^4 a_i \left(1 - \frac{T_s}{T_a}\right)^i\right), \quad (3)$$

16 where  $p_s = 101325$  Pa is the steam point pressure,  $T_s = 373.15$  K is the steam point temperature  
17 and  $a = (13.3185, -1.9760, -0.6445, -0.1299)$  are empirical coefficients (Richards, 1971).

18 The derivative of the saturated specific humidity with respect to temperature,  $\Delta$  (kg kg<sup>-1</sup> K<sup>-1</sup>),  
19 is therefore

$$20 \quad \Delta = \frac{T_s}{T_a^2} \frac{p_s q_s}{p_s - (1 - \epsilon) e_s} \sum_{i=1}^4 i a_i \left(1 - \frac{T_s}{T_a}\right)^{i-1}. \quad (4)$$

21 The available energy,  $A$  (W m<sup>-2</sup>), is the energy balance of the surface,

$$22 \quad A = R_n - G, \quad (5)$$

23 where  $R_n$  is the net radiation (W m<sup>-2</sup>) and  $G$  is the soil heat flux (W m<sup>-2</sup>). The net soil heat flux  
24 is negligible at the daily timescale (Allen et al., 1998), so the available energy is equal to the  
25 net radiation, such that

$$26 \quad A = (1 - \alpha) S_d + \epsilon (L_d - \sigma T_*^4), \quad (6)$$



1 where  $\sigma$  is the Stefan-Boltzmann constant,  $\alpha$  is the albedo and  $\epsilon$  the emissivity of the surface  
 2 and  $T^*$  is the surface temperature (Shuttleworth, 2012). For this study the surface temperature  
 3 is approximated by using the air temperature,  $T_a$ . The albedo and emissivity are also dependent  
 4 on the land cover; for a well-watered grass surface an albedo of 0.23 and an emissivity of 0.92  
 5 are used (Allen et al., 1998).

6 The air density,  $\rho_a$  ( $\text{kg m}^{-3}$ ), is a function of air pressure and temperature,

$$7 \quad \rho_a = \frac{p^*}{rT_a}, \quad (7)$$

8 where  $r = 287.05 \text{ J kg}^{-1} \text{ K}^{-1}$  is the gas constant of air.

9 The stomatal and aerodynamic resistances are strongly dependent on the land cover due to  
 10 differences in roughness length and physiological constraints on transpiration of different  
 11 vegetation types. As a standard, the Food and Agriculture Organization of the United Nations  
 12 (FAO) recommend the use of PET calculated for a hypothetical reference crop, which  
 13 corresponds to a well-watered grass crop of 0.12 m height, with constant stomatal resistance,  $r_s$   
 14 =  $70.0 \text{ s m}^{-1}$  (Allen et al., 1998). Following this recommendation, the PET in the current study  
 15 was calculated for the reference crop over the whole of Great Britain. If necessary, this can be  
 16 adjusted to give an estimate of PET specific to the local land cover, for example using  
 17 regression relationships (Crooks and Naden, 2007).

18 Following Allen et al. (1998), aerodynamic resistance,  $r_a$  ( $\text{s m}^{-1}$ ), is a function of the 10 m wind  
 19 speed

$$20 \quad r_a = \frac{278}{u_{10}}. \quad (8)$$

21 Thus the PET is a function of six of the meteorological variables: air temperature, specific  
 22 humidity, downward long- and shortwave radiation and surface air pressure.

23 The PET can be split between two factors, the radiative component,  $E_{PR}$ ,

$$24 \quad E_{PR} = \frac{t_d}{\lambda} \frac{\Delta A}{\Delta + \gamma \left(1 - \frac{r_s}{r_a}\right)}, \quad (9)$$

25 and the aerodynamic component,  $E_{PA}$ ,

$$26 \quad E_{PA} = \frac{c_p \rho_a}{\lambda} \frac{r_a (q_s - q_a)}{\Delta + \gamma \left(1 - \frac{r_s}{r_a}\right)}, \quad (10)$$

27 such that  $E_P = E_{PR} + E_{PA}$ .



### 1 3.1 Potential evapotranspiration plus interception (PETI)

2 When rain falls, water is intercepted by the canopy. The evaporation of this intercepted water  
3 is subject to the same aerodynamic resistance, defined by the roughness of the canopy, as  
4 transpiration, but is not constrained by stomatal resistance (Shuttleworth, 2012). At the same  
5 time, leaves covered with water cannot transpire, so interception inhibits transpiration from the  
6 wet fraction of the canopy (Ward and Robinson, 2000). In the short term after a rain event,  
7 potential water losses due to evaporation may be underestimated if only potential transpiration  
8 is calculated. This can be accounted for by introducing an interception term to the calculation  
9 of PET. If the daily rainfall is greater than zero, then the rain is used to fill (part of) the canopy  
10 and this store evaporates as interception, inhibiting the transpiration. On days without rain, the  
11 potential is equal to the PET defined in Eq. 2. A similar correction is applied to the PET  
12 provided at 40 km resolution by MORECS (Thompson et al., 1981).

13 The potential evapotranspiration plus interception (PETI) is a function of the PET,  $E_P$ , (as  
14 calculated above) and potential interception,  $E_I$ , which is calculated by substituting  $r_s=0 \text{ s m}^{-1}$   
15 into Eq. (2). To calculate the relative proportions of interception and transpiration, it is assumed  
16 that the wet fraction of the canopy is proportional to the amount of water in the interception  
17 store and that transpiration is only possible through the fraction of the canopy which is dry. The  
18 interception store,  $S_I$  ( $\text{kg m}^{-2}$ ), decreases through the day according to an exponential dry down  
19 (Rutter et al., 1971), such that

$$20 \quad S_I(t) = S_0 e^{-\frac{E_I}{S_{tot}}t}, \quad (11)$$

21 where  $E_I$  is the potential interception,  $S_{tot}$  is the total capacity of the interception store ( $\text{kg m}^{-2}$ ),  
22  $S_0$  is the precipitation that is intercepted by the canopy ( $\text{kg m}^{-2}$ ) and  $t$  is the time (in days) since  
23 a rain event. The total capacity of the interception store is calculated following Best et al.  
24 (2011), such that

$$25 \quad S_{tot} = 0.5 + 0.05\Lambda, \quad (12)$$

26 where  $\Lambda$  is the leaf area index (LAI); for the FAO standard grass land cover the LAI is 2.88  
27 (Allen et al., 1998). The fraction of rainfall intercepted by the canopy is found also following  
28 Best et al. (2011), assuming that rainfall lasts for an average of 3 hours.

29 The wet fraction of the canopy,  $C_{wet}$ , is proportional to the store size, such that

$$30 \quad C_{wet}(t) = \frac{S(t)}{S_{tot}}. \quad (13)$$



1 The total PETI is the sum of the interception from the wet canopy and the transpiration from  
2 the dry canopy,

$$3 \quad E_{PI}(t) = E_I C_{wet}(t) + E_P(1 - C_{wet}(t)). \quad (14)$$

4 This is integrated over one day to find the total PETI,  $E_{PI}$  (mm d<sup>-1</sup>), to be

$$5 \quad E_{PI} = S_0 \left(1 - e^{-\frac{E_I}{S_{tot}}}\right) + E_P \left(1 - \frac{S_0}{E_I} \left(1 - e^{-\frac{E_I}{S_{tot}}}\right)\right). \quad (15)$$

6 The PETI is a function of the same six meteorological variables as the PET, plus the  
7 precipitation.

### 8 **3.2 Spatial and seasonal patterns of potential evapotranspiration**

9 Both PET and PETI have a distinct gradient from low in the north-west to high in the south-  
10 east, and they are both inversely proportional to the elevation (Fig. 4), reflecting the spatial  
11 patterns of the meteorological variables. The PETI is higher than the PET overall but this  
12 difference is larger in the north and west, where precipitation rates, and therefore interception,  
13 are higher (Fig. 4). In Scotland, the higher interception and lower evaporative demand mean  
14 that this increase is a larger proportion of the total, with the mean PETI being 10% larger than  
15 the PET (in some areas the difference is more than 25%). In the English lowlands the difference  
16 is more moderate, at 6%, but it is a more water limited region where hydrological modelling  
17 can be sensitive to even relatively small adjustments to PET (Kay et al., 2013).

18 The seasonal climatology of both PET and PETI follow the meteorology (Fig. 5), with high  
19 values in the summer and low in the winter. The absolute difference between PET and PETI is  
20 bimodal, with a peak in March and a smaller peak in October (September in Scotland) (Fig. 5),  
21 because in winter the overall evaporative demand is low, while in summer the amount of rainfall  
22 is low, so the interception correction is small. The seasonal cycle of PET is driven  
23 predominantly by the radiative component, which has a much stronger seasonality than the  
24 aerodynamic component (Fig. 6).

25 On a monthly or annual timescale, the ratio of PET to precipitation is an indicator of the wet-  
26 or dryness of a region (Kay et al., 2013). Low values of PET relative to precipitation indicate  
27 wet regions, where evaporation is demand-limited, while high values indicate dry, water-  
28 limited regions. In the wetter regions (Scotland, Wales) mean-monthly PET and PETI (Fig. 5)  
29 are on average lower than the mean-monthly precipitation (Fig. 2) throughout the year, while



1 in drier regions (England, English lowlands) the PET and PETI are higher than the precipitation  
2 for much of the summer, highlighting the region's susceptibility to hydrological drought  
3 (Folland et al., 2015).

#### 4 **4 Decadal trends**

5 Annual means of the meteorological variables (Fig. 7) and the PET and PETI (Fig. 8) were  
6 calculated for each of the five regions. The trends in these annual means were calculated using  
7 linear regression; the significance ( $P$  value) and 95% confidence intervals of the slope are  
8 calculated assuming a non-zero lag-1 autocorrelation, to account for possible correlation  
9 between adjacent data points (von Storch and Zwiers, 1999; Zwiers and von Storch, 1995). In  
10 addition, seasonal means were calculated, with the four seasons defined to be December-  
11 February, March-May, June-August and September-November, and trends in these means were  
12 also found.

13 The trends and associated 95% confidence intervals of the annual means for Great Britain of  
14 the meteorological variables can be seen in Table 2. The trends in the annual and seasonal  
15 means for all regions are plotted in Fig. 9; trends that are statistically significant at the 5% level  
16 are plotted with solid error bars, those that are not significant are plotted with dashed lines.  
17 There was a statistically significant trend in air temperature in all regions (except in winter),  
18 which agrees with recent trends in the Hadley Centre Central England Temperature (HadCET)  
19 dataset (Parker and Horton, 2005) and in temperature records for Scotland (Jenkins et al., 2008)  
20 as well as in the CRUTEM4 dataset (Jones et al., 2012). An increase in winter precipitation in  
21 Scotland is seen in the current dataset, but no significant trends otherwise. Long term  
22 observations show that there has been little trend in annual precipitation, but a change in  
23 seasonality with wetting winters and drying summers (Jenkins et al., 2008). The statistically  
24 significant decline in wind speed in all regions is consistent with the results of McVicar et al.  
25 (2012) and Vautard et al. (2010), who report decreasing wind speeds in the northern hemisphere  
26 over the late 20<sup>th</sup> century.

27 The slopes and associated 95 % confidence intervals of PET and PETI for annual means over  
28 Great Britain can be seen in Table 2, and the trends in the annual and seasonal means of PET,  
29 PETI, and the radiative and aerodynamic components of PET are plotted in Fig. 10 for all  
30 regions. There is a statistically significant increase in annual PET in all regions except Wales;  
31 the GB trend ( $0.021 \text{ mm d}^{-1} \text{ decade}^{-1}$ ) is equivalent to an increase of  $0.11 \text{ mm d}^{-1}$  (8 % of the  
32 long term mean) over the whole dataset. Increases in PETI are only statistically significant in



1 England ( $0.023 \text{ mm d}^{-1} \text{ decade}^{-1}$ ) and English lowlands ( $0.028 \text{ mm d}^{-1} \text{ decade}^{-1}$ ), where the  
2 increases over the whole dataset are  $0.12 \text{ mm d}^{-1}$  (8% of the long term mean) and  $0.15 \text{ mm d}^{-1}$   
3 ( $10 \%$  of the long term mean) respectively. There is a difference in trend between different  
4 seasons. In winter, summer and autumn there are no statistically significant trends in PET or  
5 PETI, other than the English lowlands in autumn, but the spring is markedly different, with very  
6 significant trends ( $P < 0.0005$ ) in all regions. The GB spring trends in PET ( $0.043 \text{ mm d}^{-1} \text{ decade}^{-1}$ )  
7 and PETI ( $0.038 \text{ mm d}^{-1} \text{ decade}^{-1}$ ) are equivalent to an increase of  $0.22 \text{ mm d}^{-1}$  (14 % of the  
8 long-term spring mean) and  $0.20 \text{ mm d}^{-1}$  (11 % of the long-term spring mean) over the length  
9 of the dataset respectively. The radiative component of PET has similarly significant trends in  
10 spring, while the aerodynamic component has no significant trends in any season (Fig. 10),  
11 indicating that the trend in PET is due to the increasing radiative component.

12 There are few studies of long-term trends in evaporative demand in the UK. MORECS provides  
13 an estimate of Penman-Monteith PET calculated directly from the 40 km resolution  
14 meteorological data (Hough and Jones, 1997; Thompson et al., 1981), and increases can be seen  
15 over the dataset (Rodda and Marsh, 2011). But as the PET and PETI in the current dataset are  
16 ultimately calculated using the same meteorological data (albeit by different methods), it is not  
17 unexpected that similar trends should be seen. Site-based studies suggest an increase over recent  
18 decades (Burt and Shahgedanova, 1998; Crane and Hudson, 1997), but it is difficult to separate  
19 climate-driven trends from local land-use trends. The global review paper by (McVicar et al.,  
20 2012) identifies a trend of decreasing evaporative demand in the northern hemisphere, driven  
21 by decreasing wind speeds, however they also report significant local variations on trends in  
22 pan evaporation, including the increasing trend observed by Stanhill and Möller (2008) at a site  
23 in England after 1968. Matsoukas et al. (2011) identify a statistically significant increase in P  
24 in several regions of the globe, including southern England, between 1983 and 2008, attributing  
25 it predominantly to an increase in the radiative component of PET, due to global brightening.

26 Regional changes in actual evaporative losses can be estimated indirectly using regional  
27 precipitation and runoff or river flow. Using a combination of observations and modelling,  
28 Marsh and Dixon (2012) identified an increase in evaporative losses in Great Britain from 1961-  
29 2011. Hannaford and Buys (2012) note seasonal and regional differences in trends in observed  
30 river flow, suggesting that decreasing spring flows in the English lowlands are indicative of  
31 increasing evaporative demand. However, changing evaporative losses can also be due to



1 changing supply through precipitation, so it is important to formally attribute the trends in PET  
2 to changing climate, in order to understand changing evaporative losses.

### 3 **4.1 Attribution of trends in potential evapotranspiration**

4 In order to attribute changes in PET to changes in climate, the rate of change of PET,  $dE_p/dt$ ,  
5 can be calculated as a function of the rate of change of each variable (Donohue et al., 2010),

$$6 \frac{dE_p}{dt} = \frac{dE_p}{dT_a} \frac{dT_a}{dt} + \frac{dE_p}{dq_a} \frac{dq_a}{dt} + \frac{dE_p}{du_{10}} \frac{du_{10}}{dt} + \frac{dE_p}{dL_d} \frac{dL_d}{dt} + \frac{dE_p}{dS_d} \frac{dS_d}{dt}. \quad (16)$$

7 Note that we exclude the surface air pressure, as the interannual variability of air pressure is  
8 negligible. The derivative of the PET with respect to each of the meteorological variables can  
9 be found analytically (Appendix A). The derivatives are calculated from the daily  
10 meteorological data, then the overall annual and regional means found. Substituting the slopes  
11 of the linear regressions of the annual means (Fig. 9) for the rate of change of each variable  
12 with time, the contribution of each variable to the rate of change of PET can be calculated. The  
13 same can also be applied to the radiative and aerodynamic components independently.

14 Figure 11 shows the contribution of each meteorological variable to the rate of change of the  
15 annual mean PET and to the radiative and aerodynamic components. The percentage  
16 contribution is seen in Table 3. The radiative component has no dependence on specific  
17 humidity, while the aerodynamic component has no dependence on long- or shortwave  
18 radiation.

19 The rate of change of PET is almost entirely due to the change in the radiative component. In  
20 all regions except Scotland, the change in the radiative component of PET is dominated by the  
21 increase due to the increasing downward shortwave radiation, followed by the increasing  
22 downward longwave radiation, while in Scotland the effect of the downward shortwave is  
23 smaller. In all regions there is also a small increase in the radiative component due to the  
24 decreasing wind speed, and a decrease due to increasing air temperature, but these are negligible  
25 compared to the effect of changing radiation. Increasing air temperature contributes to a small  
26 increase in the aerodynamic component of PET, but this is offset by the decrease due to  
27 increasing specific humidity and decreasing wind speed, so that overall the change in the  
28 aerodynamic component is negligible.



## 1 **5 Discussion**

2 These high resolution datasets provide an insight into the effect of the changing climate of Great  
3 Britain on evaporative demand over the past five decades. There have been significant climatic  
4 trends in the UK since 1961; in particular rising air temperature and specific humidity,  
5 decreasing wind speed and decreasing cloudiness. The resulting trends in downward long- and  
6 shortwave radiation have combined to lead to trends in evaporative demand.

7 Wind speeds have decreased more significantly in the west than the east, and show a consistent  
8 decrease across seasons. Contrary to Donohue et al. (2010) and McVicar et al. (2012), this study  
9 finds that the change in wind speed of the late 20<sup>th</sup> and early 21<sup>st</sup> centuries has had a negligible  
10 influence on PET over the period of study. However, the previous studies were concerned with  
11 open-water Penman evaporation, which has a simpler (proportional) dependence on wind speed  
12 than the Penman-Monteith PET considered here. Although the significant decrease in wind  
13 speed has had a negligible effect on evaporative demand, it may nonetheless have had a direct  
14 effect on biodiversity (Barton, 2014; Brittain et al., 2013) and implications for wind energy  
15 resources (Sinden, 2007).

16 The air temperature trends in this study of around 0.2 K decade<sup>-1</sup> are consistent with observed  
17 global and regional trends (Hartmann et al., 2013; Jenkins et al., 2008). The temperature trend  
18 also does not explicitly make a large contribution to the trend in PET, but is partly responsible  
19 for the trend of increasing downward longwave radiation. The trends in longwave radiation in  
20 these datasets are not statistically significant, due to high inter-annual variability, but contribute  
21 to between 22% and 50% of the trends in PET and the radiative component (Table 3).  
22 Observations of longwave radiation are often uncertain, but, although small, the trend in this  
23 dataset is consistent with observed trends (Wang and Liang, 2009), as well as with trends in the  
24 WFDEI bias-corrected reanalysis product (Weedon et al., 2014).

25 Increasing solar radiation has been shown to have a strong effect on spring and annual  
26 evaporative demand, contributing to between 46% and 77% of the trend in annual PET (Table  
27 3), increasing to between 84% and 87% of the trend in spring PET. Two main mechanisms can  
28 be responsible for changing solar radiation – changing cloud cover and changing aerosol  
29 concentrations. Changing aerosol emissions have been shown to have had a significant effect  
30 on solar radiation in the 20<sup>th</sup> century. In Europe, global dimming due to increased aerosol  
31 concentrations peaked around 1980, followed by global brightening as aerosol concentrations  
32 decreased (Wild, 2009). Observations of changing continental runoff and river flow in Europe





1 over the 20<sup>th</sup> century have been attributed to changing aerosol concentrations, via their effect  
2 on solar radiation, and thus evaporative demand (Gedney et al., 2014).

3 In this study we use the duration of bright sunshine to calculate the solar radiation, using  
4 empirical coefficients which do not vary with year, so aerosol effects are not explicitly included.  
5 The coefficients used in this study to convert sunshine hours to radiation fluxes were  
6 empirically derived in 1978; the derivation used data from the decade 1966-75, as this period  
7 was identified to be before reductions in aerosol emissions had begun to significantly increase  
8 observed solar radiation (Cowley, 1978). Despite this, the trend in shortwave radiation in the  
9 current dataset from 1979 onwards is consistent, within uncertainties, with that seen in the  
10 WFDEI data, which is bias-corrected to observations and includes explicit aerosol effects  
11 (Weedon et al., 2014).

12 It has been suggested that aerosol effects also implicitly affect sunshine duration (Helmes and  
13 Jaenicke, 1986). Several regional studies have shown trends in sunshine hours that are  
14 consistent with the periods of dimming and brightening across the globe (eg Liley, 2009;  
15 Sanchez-Lorenzo et al., 2009; Sanchez-Lorenzo et al., 2008; Stanhill and Cohen, 2005), and  
16 several have attempted to quantify the relative contribution of trends in cloud cover and aerosol  
17 loading (e.g. Sanchez-Lorenzo and Wild (2012) in Switzerland, see Sanchez-Romero et al.  
18 (2014) for a review). Therefore, it may be that some of the brightening trend seen in the current  
19 dataset is due to the implicit signal of aerosol trends in the MORECS sunshine duration,  
20 although this is likely to be small compared to the effects of changing cloud cover.

21 The trends in the MORECS sunshine duration used in this study are consistent with changing  
22 weather patterns which may be attributed to the Atlantic Multidecadal Oscillation (AMO). The  
23 AMO has been shown to cause a decrease in spring precipitation (and therefore cloud cover) in  
24 northern Europe over recent decades (Sutton and Dong, 2012), and the trend in MORECS  
25 sunshine hours is dominated by an increase in the spring mean. This has also been seen in  
26 Europe-wide sunshine hours data (Sanchez-Lorenzo et al., 2008). On the other hand, the effect  
27 of changing aerosols on sunshine hours is expected to be largest in the winter (Sanchez-Lorenzo  
28 et al., 2008). However, it would not be possible to directly identify either of these effects on the  
29 sunshine duration without access to longer data records.

30 The inclusion of explicit aerosol effects in the coefficients of the Angstrom-PreScott equation  
31 would be expected to mitigate the trend in evaporative demand in the first two decades of the  
32 dataset, and enhance it after 1980. Gedney et al. (2014) attribute a decrease in European solar



1 radiation of  $10 \text{ W m}^{-2}$  between the periods 1901-10 and 1974-80, and an increase of  $4 \text{ W m}^{-2}$   
2 from 1974-84 to 1990-99 to changing aerosol contributions. Applying these trends to the current  
3 dataset, with a turning point at 1980, would double the overall increase in solar radiation in  
4 Great Britain, which would lead to a 50 % increase in the overall trend in PET.

5 The trends in temperature and cloud cover in the UK are expected to continue into the coming  
6 decades, with precipitation expected to increase in the winter but decrease in the summer  
7 (Murphy et al., 2009). Therefore it is likely that evaporative demand will increase, increasing  
8 water stress in the summer when precipitation is lower and potentially affecting water resources,  
9 agriculture and biodiversity. This has been demonstrated for southern England and Wales by  
10 Rudd and Kay (2015), who calculated present and future PET using high-resolution RCM  
11 output and include  $\text{CO}_2$  fertilisation.

12 The current study is concerned only with the effects of changing climate on evaporative demand  
13 and has assumed a constant bulk canopy resistance throughout. However, plants are expected  
14 to react to increased  $\text{CO}_2$  in the atmosphere by closing stomata and limiting the exchange of  
15 gases, including water (Kruijt et al., 2008), and observed changes in runoff have been attributed  
16 to this effect (Gedney et al., 2006; Gedney et al., 2014). It is possible that the resulting change  
17 of canopy resistance could partially offset the increased atmospheric demand (Rudd and Kay,  
18 2015) and may impact runoff (Gedney et al., 2006; Prudhomme et al., 2014), but further studies  
19 would be required to quantify this.

20 This paper has presented a unique high-resolution observation-based dataset of meteorological  
21 variables and evaporative demand in Great Britain since 1961. We have shown that trends in  
22 evaporative demand can be attributed to trends in the meteorological variables. The  
23 meteorological variables provided are sufficient to run land surface models and combined with  
24 the PET can be used to run hydrological models. In addition, the high spatial (1km) and  
25 temporal (daily) resolution will allow this dataset to be used to study the effects of climate on  
26 physical and biological systems at a range of scales, from local to national.

### 27 **Author contribution**

28 EB, JF and DBC designed the study. JF, ACR, DBC and ELR developed code to create  
29 meteorological data. ELR created the PET and PETI. ELR and EB analysed trends. ELR, EB,  
30 ACR and DBC wrote the manuscript.

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1 **Appendix A: Derivatives of potential evapotranspiration**

2 The wind speed affects the PET through the aerodynamic resistance. The derivative with respect  
 3 to wind speed is

$$4 \quad \frac{\partial E_P}{\partial u_{10}} = \frac{(\Delta + \gamma)E_{PA} - \gamma \frac{r_s}{r_a} E_{PR}}{u_{10} \left( \Delta + \gamma \left( 1 + \frac{r_s}{r_a} \right) \right)}. \quad (A1)$$

5 The downward long- and shortwave radiation affect PET through the net radiation, and the  
 6 derivatives are

$$7 \quad \frac{\partial E_P}{\partial L_d} = E_{PR} \frac{\epsilon}{R_n} \quad (A2)$$

$$8 \quad \frac{\partial E_P}{\partial S_d} = E_{PR} \frac{(1 - \alpha)}{R_n}. \quad (A3)$$

9 The derivative of PET with respect to specific humidity is

$$10 \quad \frac{\partial E_P}{\partial q_a} = \frac{E_{PA}}{q_a - q_s}. \quad (A4)$$

11 The air temperature affects PET through the saturated specific humidity and its derivative, the  
 12 net radiation and the air density, so that the derivative of PET with respect to air temperature is

$$13 \quad \frac{\partial E_P}{\partial T_a} = E_{PR} \left( \frac{\gamma \left( 1 + \frac{r_s}{r_a} \right)}{T_a^2 \left( \Delta + \gamma \left( 1 + \frac{r_s}{r_a} \right) \right)} \left[ T_{sp} \left( \sum_{i=1}^4 i a_i T_r^{i-1} + \frac{\sum_{i=1}^4 i a_i T_r^{i-1}}{\sum_{i=1}^4 i(i-1) a_i T_r^{i-2}} + \frac{2(1-\epsilon) \sum_{i=1}^4 i a_i T_r^{i-1} q_s}{\epsilon} \right) - \right. \right.$$

$$14 \quad \left. \left. 2T_a \right] - \frac{4\epsilon\sigma T_a^4}{R_n} \right) + E_{PA} \left( \frac{\Delta}{q_s - q} - \frac{1}{T_a} - \frac{\Delta}{T_a^2 \left( \Delta + \gamma \left( 1 + \frac{r_s}{r_a} \right) \right)} \left[ T_{sp} \left( \sum_{i=1}^4 i a_i T_r^{i-1} + \frac{\sum_{i=1}^4 i a_i T_r^{i-1}}{\sum_{i=1}^4 i(i-1) a_i T_r^{i-2}} + \right. \right. \right.$$

$$15 \quad \left. \left. \frac{2(1-\epsilon) \sum_{i=1}^4 i a_i T_r^{i-1} q_s}{\epsilon} \right) - 2T_a \right] \right). \quad (A5)$$



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1 Table 1. Variable details

Variable (units)	Source data	Ancillary files	Assumptions	Height
Air temperature (K)	MORECS air temperature	IHDTM elevation	Lapsed to IHDTM elevation	1.2 m
Specific humidity (kg kg <sup>-1</sup> )	MORECS vapour pressure, air temperature	IHDTM elevation	Lapsed to IHDTM elevation  Constant air pressure	1.2 m
Downward longwave radiation (W m <sup>-2</sup> )	MORECS air temperature, vapour pressure, sunshine hours	IHDTM elevation	Constant cloud base height	n/a
Downward shortwave radiation (W m <sup>-2</sup> )	MORECS sunshine hours	IHDTM elevation  Spatially-varying aerosol correction	No time-varying aerosol correction	n/a
Wind speed (m s <sup>-1</sup> )	MORECS wind speed	ETSU average wind speeds	Wind speed correction is constant	10 m
Precipitation (kg m <sup>-2</sup> s <sup>-1</sup> )	CEH-GEAR precipitation		No transformations performed	n/a
Daily temperature range (K)	CRU TS 3.21 daily temperature range		No spatial interpolation from 0.5° resolution.  No temporal interpolation (constant values for each month)	1.2 m



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Surface air pressure (Pa)	WFD air pressure	IHDTM elevation	Mean-monthly values from WFD used (each year has same values). Lapsed to IHDTM elevation. No temporal interpolation (constant values for each month).	n/a
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- 1 Table 2: Rate of change of annual means of meteorological and potential evapotranspiration  
 2 variables in Great Britain. Bold indicates trends that are significant at the 5% level. Numbers  
 3 in brackets show the 95% confidence intervals.

Variable	Rate of change (95% confidence interval)
Air temperature	<b>0.20 (0.07, 0.31) K decade<sup>-1</sup></b>
Specific humidity	<b>0.046 (0.010, 0.082) g kg<sup>-1</sup> decade<sup>-1</sup></b>
Downward shortwave radiation	<b>1.1 (0.3, 1.8) W m<sup>-2</sup> decade<sup>-1</sup></b>
Downward longwave radiation	0.45 (-0.01,0.91) W m <sup>-2</sup> decade <sup>-1</sup>
Wind speed	-0.17 (-0.27, -0.08) m s <sup>-1</sup> decade <sup>-1</sup>
Precipitation	<b>0.08 (0.02, 0.14) mm day<sup>-1</sup> decade<sup>-1</sup></b>
Daily temperature range	-0.06 (-0.12,0.00) K decade <sup>-1</sup>
PET	<b>0.021 (0.00,0.041) mm day<sup>-1</sup> decade<sup>-1</sup></b>
PETI	0.019 (0.00,0.039) mm day <sup>-1</sup> decade <sup>-1</sup>

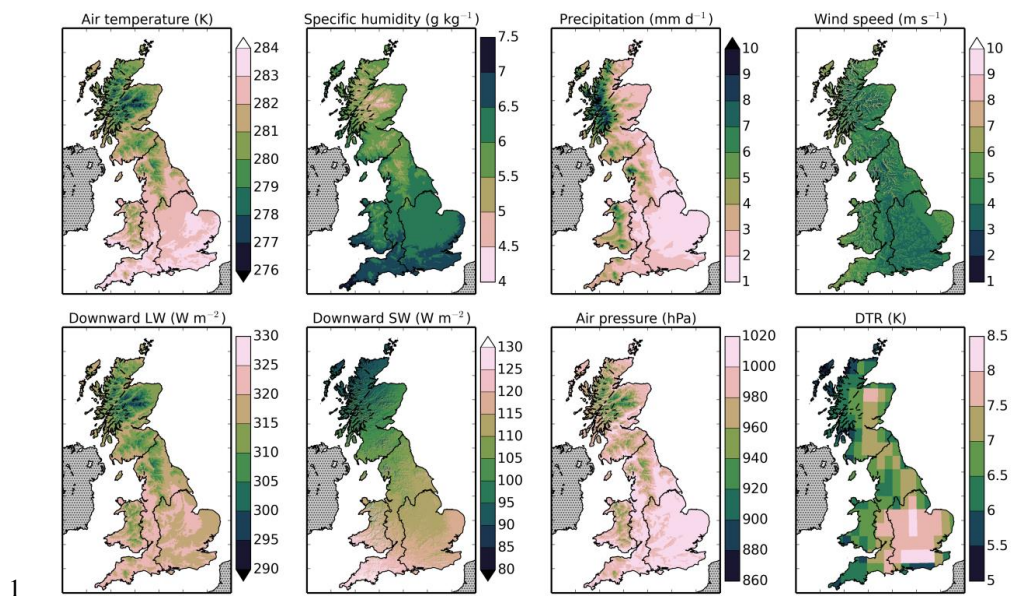
4



- 1 Table 3. Percentage contribution of the trend in each variable to the trends in annual mean PET  
2 and its radiative and aerodynamic components.

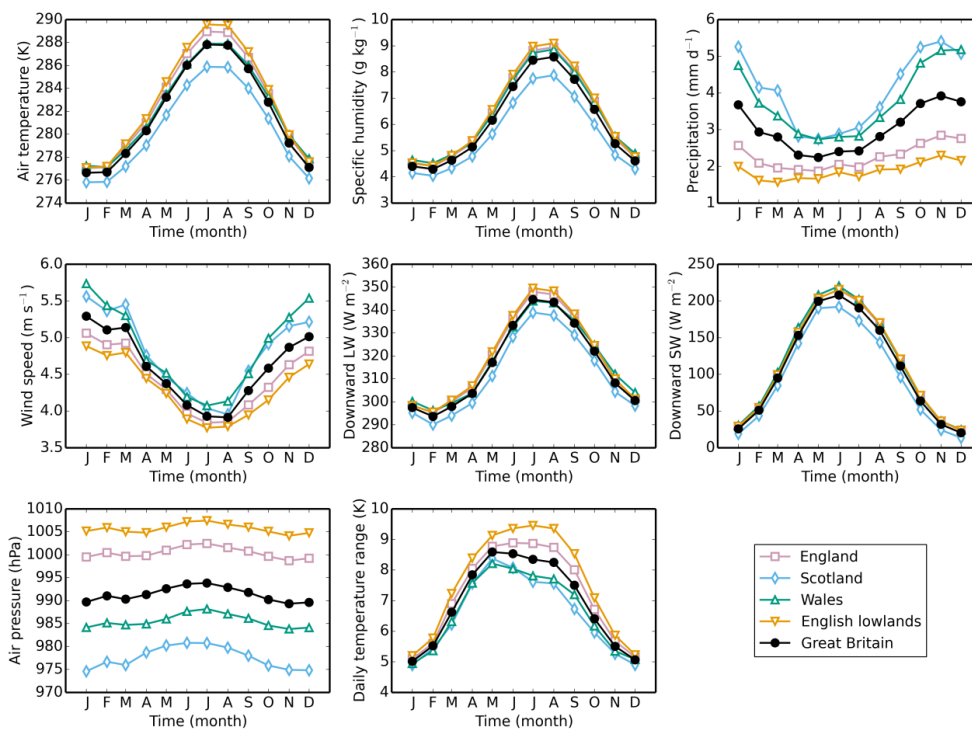
a) Potential evapotranspiration (PET)					
	Air temperature	Specific humidity	Wind speed	Downward longwave	Downward shortwave
England	7.7 %	-4.6 %	-1.8 %	26.4 %	72.3 %
Scotland	9.2 %	-6.0 %	-3.2 %	53.4 %	46.5 %
Wales	8.2 %	-5.6 %	-2.4 %	32.7 %	67.0 %
English lowlands	7.3 %	-4.0 %	-1.4 %	22.7 %	75.3 %
Great Britain	8.1 %	-5.1 %	-2.2 %	33.9 %	65.3 %
b) Radiative component of PET					
	Air temperature	Specific humidity	Wind speed	Downward longwave	Downward shortwave
England	-1.6 %	n/a	1.5 %	26.8 %	73.3 %
Scotland	-1.9 %	n/a	2.5 %	53.1 %	46.3 %
Wales	-1.5 %	n/a	2.8 %	32.3 %	66.3 %
English lowlands	-1.7 %	n/a	1.1 %	23.3 %	77.2 %
Great Britain	-1.7 %	n/a	1.9 %	34.1 %	65.7 %
c) Aerodynamic component of PET					
	Air temperature	Specific humidity	Wind speed	Downward longwave	Downward shortwave
England	703.7 %	-353.5 %	-250.2 %	n/a	n/a
Scotland	-1210.0 %	662.2 %	647.3 %	n/a	n/a
Wales	-854.7 %	492.3 %	462.5 %	n/a	n/a
English lowlands	365.4 %	-165.8 %	-99.6 %	n/a	n/a
Great Britain	2025.0 %	-1061.9 %	-863.1 %	n/a	n/a

3



2 Figure 1. Means of the meteorological variables over the years 1961-2012. Top row, left to  
3 right are 1.2 m air temperature, 1.2 m specific humidity, precipitation, 10 m wind speed.  
4 Bottom row left to right are downward longwave radiation, downward shortwave radiation,  
5 surface air pressure, daily air temperature range.



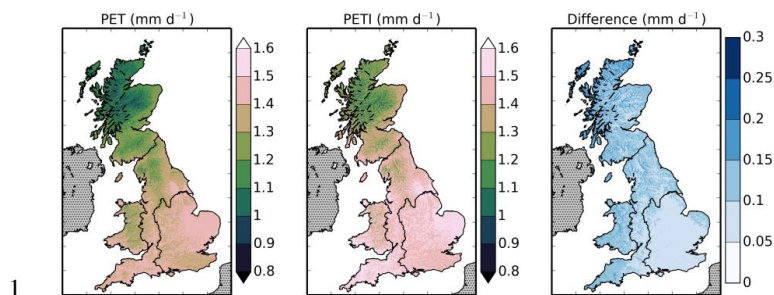


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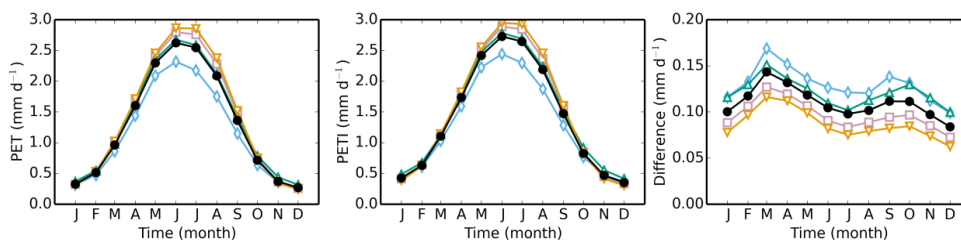
2 Figure 2. Mean monthly climatology of meteorological variables for five different regions of  
 3 Great Britain, calculated over the years 1961-2012.



1  
2 Figure 3. The regions used to calculate the area means. The English lowlands are a sub-region  
3 of England. England, Scotland and Wales together form the fifth region, Great Britain.

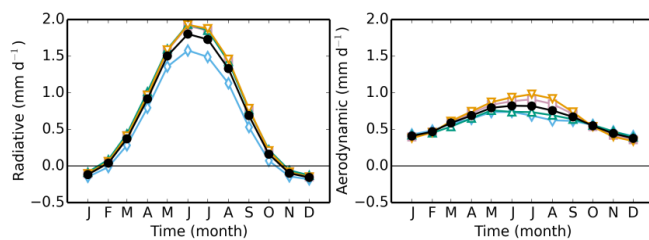


2 Figure 4. Mean PET (right), mean PETI (centre), and the difference between mean PETI and  
3 PET (right), calculated over the years 1961-2012.



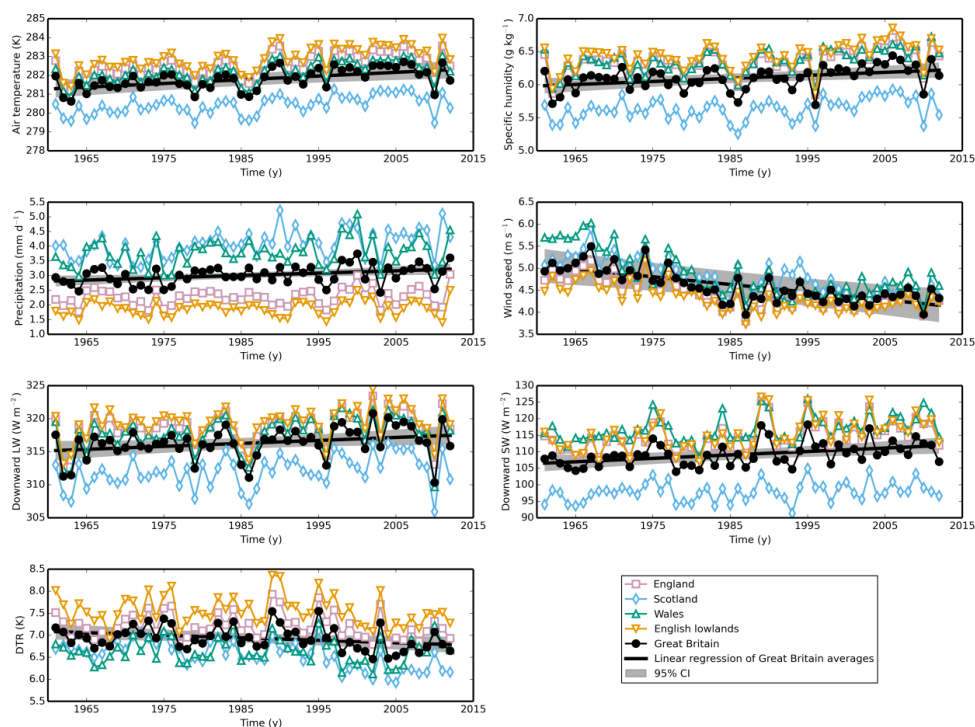
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2 Figure 5. Mean monthly climatology of PET (left), PETI (centre) and the difference PETI-PET  
3 (right) for five different regions of Great Britain, calculated over the years 1961-2012. Symbols  
4 as in Fig. 2.



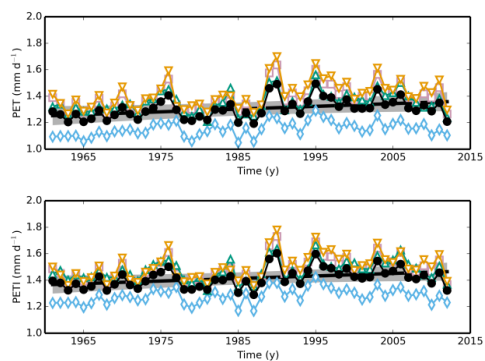
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- 2 Figure 6. Mean-monthly climatology of the radiative (left) and aerodynamic (right) components  
3 of the PET for five different regions of Great Britain, calculated over the years 1961-2012.  
4 Symbols as in Fig. 2.



1

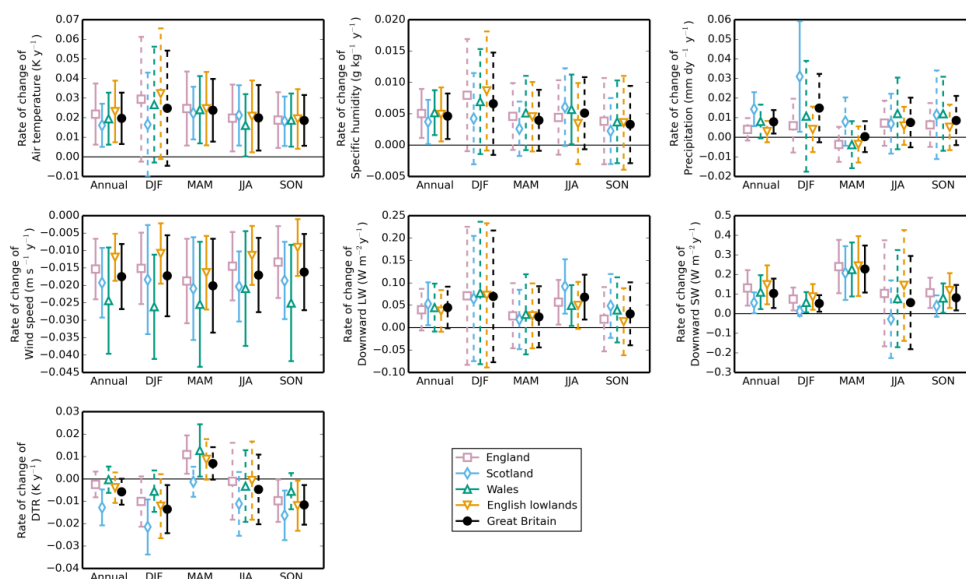
2 Figure 7. Annual means of the meteorological variables over five regions of Great Britain. The  
3 solid black lines show the linear regression fit to the Great Britain annual means, while the grey  
4 strip shows the 95% confidence interval of the same fit, assuming a non-zero lag-1 correlation  
5 coefficient.



1

2 Figure 8. Annual means of PET and PETI for five regions of Great Britain. Symbols as in Fig.

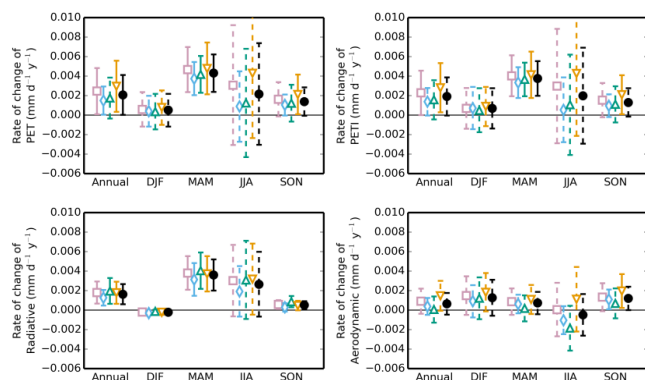
3 7.



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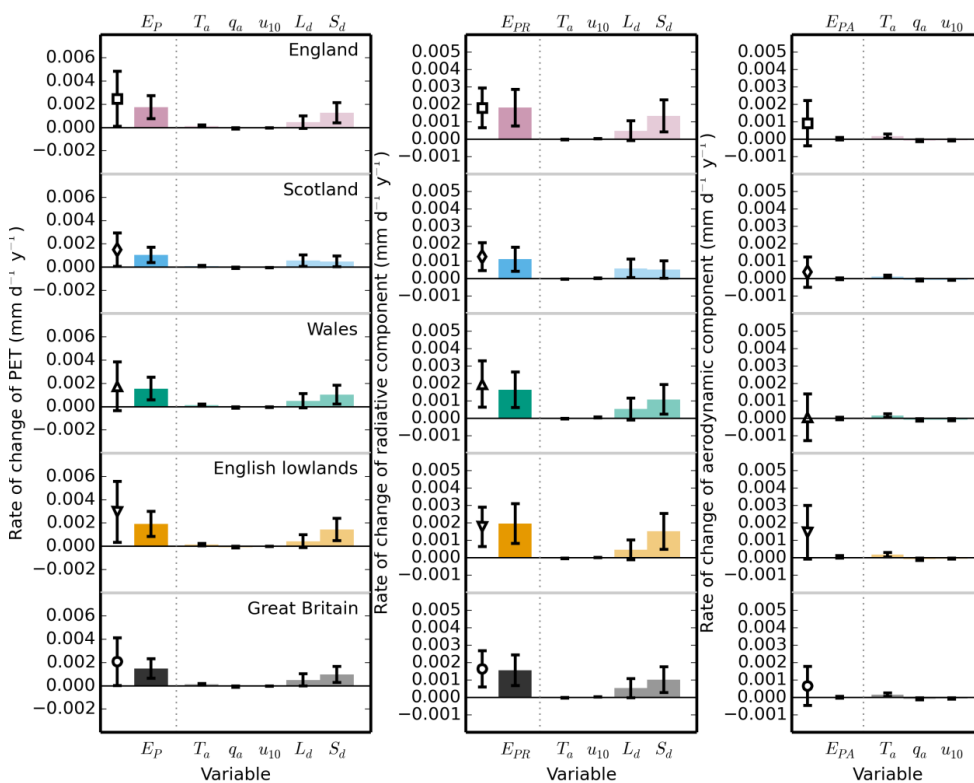
2 Figure 9. Rate of change of annual and seasonal means of meteorological variables for five  
3 regions of Great Britain for the years 1961-2012. Error bars are the 95% confidence intervals  
4 calculated assuming a non-zero lag-1 correlation coefficient. Solid error bars indicate slopes  
5 that are statistically significant at the 5% level, dashed error bars indicate slopes that are not  
6 significant at the 5% level.





1

2 Figure 10. Rate of change of annual and seasonal means of PET (top left), PETI (top right), the  
3 radiative component of PET (lower left) and the aerodynamic component of PET (lower right)  
4 for five regions of Great Britain for the years 1961-2012. Symbols as in Fig. 9.



1

2 Figure 11. The contribution of the rate of change of each meteorological variable to the rate of  
 3 change of PET (left), the radiative component (centre) and the aerodynamic component (right).  
 4 In each panel the left hand bar is the rate of change of PET derived from the rate of change of  
 5 each of the variables. The rest of the columns show the contribution to that change from each  
 6 of the variables. The error bars show the 95% confidence intervals on each value. For the left  
 7 hand bar, the symbols with error bars show the slope and its associated confidence interval  
 8 obtained from the linear regression (as in Fig. 10).