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DOI:

[10.1177/0309133314538725](https://doi.org/10.1177/0309133314538725)

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Document Version

Peer reviewed version

Citation for published version (Harvard):

Zhang, F, Cai, X & Thornes, J 2014, 'Birmingham's air and surface urban heat islands associated with Lamb weather types and cloudless anticyclonic conditions', *Progress in Physical Geography*, vol. 38, no. 4, pp. 431-447. <https://doi.org/10.1177/0309133314538725>

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Birmingham's air and surface urban heat islands associated with Lamb weather types and cloudless anticyclonic conditions

Abstract

This study investigates the characteristics of the air and surface urban heat islands (aUHI and sUHI) of Birmingham in relation to Lamb weather types (LWTs) over the period 2002-2007, with a particular focus on cloudless anticyclonic conditions. Ground-based MIDAS air temperatures within urban canopy layer at the urban Edgbaston and rural Shawbury weather stations were used to derive the aUHI intensity (aUHII). Satellite-derived MODIS/Aqua land surface temperatures (LST) under cloudless conditions were used to derive the spatial patterns of the sUHI as well as the sUHI intensity (sUHII). Using Jenkinson's objective daily synoptic indices, a combined subset of eleven LWTs were examined for their association with the nocturnal aUHII. Over the study period, the most frequently occurring LWT, 'anticyclonic' (21.1%), gives a strongest mean/maximum nocturnal aUHII of 2.5 °C/7 °C (391 nights) and the largest proportion of nocturnal heat island events of 65.2%. The spatial patterns of nocturnal sUHI for each LWT were also assessed, and the results showed that Birmingham's sUHI spatial patterns demonstrate Birmingham's urban warming of up to 4.16 °C (48 clear nights) in the city centre under cloudless anticyclonic conditions. The scatter plot of nocturnal aUHII and sUHII for the 48 nights demonstrates a linear aUHI-sUHI relationship. We also developed a simple analytical model that links the slope of the aUHI-sUHI relationship to the difference of "built-up" area fraction between the urban pixel and the rural pixel. This partially explains the physical basis behind the relationship. These findings of aUHII-sUHII relationships may lead to a future development of a generic methodology of deriving the spatial patterns of aUHI from satellite measurement.

Keywords: Lamb weather types; anticyclonic; urban heat island; air temperature; MODIS; land surface temperature; Birmingham

1 **1. Introduction**

2 During the 20th century, the urban population of England increased from 77% to 89% (Hicks and
3 Allen, 1999). With rapid urbanisation and climate change (IPCC, 2007) increased urban warming is
4 likely to trigger more issues of urban health and wellbeing. Due to the thermal properties of
5 construction materials, lack of vegetation, ‘urban canyons’ and anthropogenic heat (Smith et al.,
6 2009), the excess heat absorbed during the daytime causes warmer nights, contributing to higher
7 urban temperatures than those in surrounding rural areas. This phenomenon is referred to as the
8 urban heat island (UHI) effect, which generally refers to warmer air temperatures, but refers to
9 warmer surface temperatures as well (Voogt, 2000). It is well known that the UHI is more prominent
10 on calm and clear summer nights (Oke, 1987). The UHI impacts can be beneficial, particularly the
11 reduced winter mortality and reduced road salting due to less icy roads (Voogt, 2000). However, the
12 summer UHI in warmer cities is generally adverse to human health as it can cause additional heat
13 stress in hot weather. The significance of the UHI on heatwaves for cities was highlighted in August
14 2003 when London’s UHI intensity reached up to 6 to 9 °C, causing nearly 600 excess deaths (Mayor
15 of London, 2006). In Birmingham, there was a significant UHI effect of over 4.5 °C during the 2006
16 heatwave (Tomlinson et al., 2012a).

17 To identify the UHI, previous studies have used either direct or indirect measurement methods. Air
18 urban heat island (aUHI) can be quantified from air temperature (T_{air}) directly measured by weather
19 monitoring stations (within the urban canopy layer). Long-term trends in London’s aUHI have been
20 examined by Lee (1992), Jones and Lister (2009) and Wilby et al. (2011). Wilby (2003) projected
21 London’s future UHI using the Met Office HadCM3 climate model. Unfortunately, the UKCP09
22 projections are not able to predict the future UHI due to the missing contribution from urban surfaces
23 in regional climate models (Wilby et al., 2009). Surface urban heat island (sUHI) can be quantified
24 using land surface temperatures (LST) indirectly measured by satellite remote-sensing. Remote-
25 sensing measures radiances in various wavelength bands that are used to estimate surface
26 temperatures. The satellite images do derived for clear-sky conditions capture the spatial views of
27 urban surface temperature (i.e. the skin temperature). Satellite-derived studies have been limited in
28 low-resolution sensors such as 1.1 km pixels for Advanced Very High Resolution Radiometer
29 (AVHRR) (Roth et al., 1989) and 1 km for the Moderate Resolution Imaging Spectroradiometer
30 (MODIS) (Wan et al., 2004). There have been very few thermal studies with high spatial resolution
31 using Landsat-5 Thematic Mapper (TM) (Liu and Zhang, 2011) and Landsat-7 Enhanced Thematic
32 Mapper Plus (ETM+) (120 m for TM and 60 m for ETM+) due to limited temporal coverage and the
33 infrequent cloudless weather conditions. In recent years, the MODIS instrument operating on the
34 National Aeronautics and Space Administration (NASA)’s Aqua and Terra satellite sensors has been

1 widely used for LST measurements and sUHI studies (Hung et al., 2006). Both satellites can view the
2 entire surface of the Earth every 1-2 days and acquire data in 36 spectral bands at three spatial
3 resolutions of 250 m (Band 1-2), 500 m (Band 3-7) and 1000 m (Band 8-36) (MODIS,
4 <http://modis.gsfc.nasa.gov/>). In addition, not all bands are available at the three spatial resolutions.
5 The MODIS LST at 5 km resolution is retrieved in bands 20, 22, 23, 29 and 31-33. For the clear sky
6 pixels, the LST is retrieved in bands 31 and 32 at 1 km resolution (Wan et al., 2004) with the
7 generalized split-window algorithm (Wan and Dozier 1996). For a given time and location, different
8 satellite images will be generated using Aqua- and Terra-MODIS sensors based on their different
9 orbits. Tomlinson et al. (2012a) derived Birmingham's summer nocturnal sUHI for different
10 Pasquill-Gifford stability classes (Pasquill and Smith, 1983) using MODIS/Aqua LST data. The
11 specific case study for 2006 heatwave event found a Birmingham sUHI peak of 4.88 °C.

12 Lamb weather types (Lamb, 1950, 1972), LWTs, have been used to classify synoptic airflow
13 circulations based on the pressure patterns for the British Isles. There have been very few UHI
14 studies undertaken in relation to synoptic weather types such as that by Unwin (1980) for
15 Birmingham, showing that mean nocturnal aUHI peaked at 1.34 K in autumn (10 September-19
16 November) during 1965-1974. Anticyclonic circulation was associated with a maximum mean
17 nocturnal aUHI intensity of 2.26 K, and cyclonic circulation was associated with a minimum mean
18 nocturnal aUHI intensity of 0.49 K. Morris and Simmonds (2000) examined the relationship between
19 the UHI magnitude and anomalous synoptic conditions in Melbourne, Australia. These studies
20 suggested that anticyclonic conditions favour the development of the aUHI (and by inference the
21 sUHI). In addition, it should be noted that the frequency of LWTs vary with location, for example,
22 London is more dominated by anticyclonic, easterly and southerly weather types than Birmingham
23 due to its relatively continental location (O'Hare et al., 2005).

24 There have been some studies into the comparison of satellite-derived LST with ground-based T_{air}
25 (Hachem et al., 2012; Tomlinson et al., 2012b; White-Newsome et al., 2013; Jin and Dickinson,
26 2010), but very few on the comparison of aUHI and sUHI. Considering the lack of combined
27 research of aUHI and sUHI in relation to LWTs, this study builds upon Unwin (1980)'s synoptic
28 methods to examine Birmingham's nocturnal aUHI and follows the technical methods of Tomlinson
29 et al. (2012a) using MODIS/Aqua images to assess Birmingham's nocturnal sUHI. The authors also
30 notice the scarce of investigation of the aUHII-sUHII relationship, which is practically important
31 because this relationship will pave the way for the development of a generic methodology of deriving
32 the spatial patterns of aUHI from satellite measurement. This relationship will be examined in this
33 study for cloudless anticyclonic nights in Birmingham.

34

1 **2. Data and methods**

2 **2.1 Study area**

3 The West Midlands region is located in western central England. It contains the UK's second most
4 populous city (Birmingham), and a county, which includes seven metropolitan boroughs. It covers
5 13,004 km² with a population of 5,602,000. Birmingham covers 268 km² with a population of
6 1,073,000 (ONS, 2011). The UK Met Office is trying to select ideal stations which could represent
7 wider area around the station and not unduly affected by local effects i.e. ground level station
8 without vegetation and buildings (Met Office, 2011). Indeed, it is impossible to find a perfect
9 weather station. In this study, two weather stations, Edgbaston and Shawbury (Figure 1), located in
10 the West Midlands were selected as the urban and rural stations, respectively. The Edgbaston station
11 (52.476 N, -1.934°W) is located in Birmingham's central area approximately 2 km from the city
12 centre at an elevation of 160 m. The Shawbury station (52.795 N, -2.665 °W) is located
13 approximately 65 km to north-west of Edgbaston at an elevation of 72 m. The selection of
14 urban/rural stations is mainly based on their locations and surrounding environments (see Section 2.2,
15 Figure 2). The effects of elevation difference on the UHI calculation have been discussed by Jones
16 and Lister (2009) and Peterson and Owen (2005). Here, the 88 m elevation difference between the
17 Edgbaston and Shawbury stations is not taken directly into account in this study, but will be
18 mentioned where appropriate. In addition, the Coleshill station located approximately 4.5 km to
19 Birmingham's eastern edge is only used for obtaining the cloud cover data, which are not available at
20 the Edgbaston station.

21

22 **2.2 Land surface temperatures**

23 The MODIS sensor is carried on both Aqua and Terra satellite platforms. Nocturnal satellite images
24 obtained from the MODIS/Aqua satellite sensor pass over Birmingham at approximately 01:30 UTC
25 local time, but at approximately 10:30 UTC for Terra (Tomlinson et al., 2012a). The MODIS product,
26 *MYD11A1–MODIS/Aqua Land Surface Temperature/Emissivity (LST/E) Daily L3 Global 1 km SIN*
27 *Grid V005* was used in this study. This dataset is available from 8th July 2002 up to the present with
28 1 km spatial resolution. Indeed, the 1 km grid is more precisely 0.93 km due to the errors caused by
29 geo-location or satellite sensor observations (Wan, 2007). The data were obtained through the NASA
30 Land Processes Distributed Active Archive Centre (LP DAAC) Data Pool. According to the MODIS
31 Sinusoidal grid tiling system, the tile identifier of h17v03 was chosen to acquire the data across
32 Birmingham.

1 The 1 km² pixels covering the urban Edgbaston and rural Shawbury are shown in Figure 2 and the
2 locations of the weather stations are also marked. The Edgbaston station is surrounded by urban and
3 built-up areas, mainly covered with buildings, vegetation (trees and parks) and a small part of a
4 reservoir. By contrast, the Shawbury station is surrounded by a large proportion of grassland (or
5 arable land) with a few clusters of low density buildings.

6

7 **2.3 Air temperatures**

8 Ground-based meteorological data comprised of hourly T_{air} at the Edgbaston and Shawbury stations,
9 and cloud cover at the Coleshill station were obtained from the Met Office Integrated Data Archive
10 System (MIDAS) Land Surface Stations data held at the British Atmospheric Data Centre (BADC).
11 The data were extracted between 8th July 2002 and 31st July 2007 (2002-2007), coinciding with the
12 period of data availability of the MODIS LST. In addition, the MIDAS T_{air} is measured at the screen
13 height, and for the Edgbaston (urban) station, T_{air} is measured within the urban canopy layer. Daily
14 maximum and minimum air temperature ($T_{\text{air}/\text{max}}$ and $T_{\text{air}/\text{min}}$) were derived from the hourly T_{air} ; thus
15 the time at which $T_{\text{air}/\text{max}}$ (or $T_{\text{air}/\text{min}}$) occurred may vary from day to day. In Section 3.3.1 where T_{air}
16 is compared with LST, according to the acquisition time of satellite images at about 01:30 UTC, the
17 meteorological variables averaged at 01:00 UTC and 02:00 UTC were used.

18

19 **2.4 Lamb weather types**

20 Lamb (1972) originally classified daily synoptic weather conditions based on the variations in
21 surface pressures over the British Isles (50 - 60 °N, 2 °E - 10 °W). The following classifications of
22 LWTs are included in the analysis: eight directional types (South-easterly [SE], Southerly [S],
23 Northerly [N], Westerly [W], South-westerly [SW], North-westerly [NW], Easterly [E] and North-
24 easterly [NE]), two non-directional types (Anticyclonic [A], Cyclonic [C]), and one Unclassified [U]
25 type which represents days with weak or chaotic circulation patterns (Hulme and Barrow, 1997).
26 Jenkinson and Collinson (1977) improved an objective LWTs using daily grid point mean sea-level
27 pressure data, to classify daily circulation patterns into twenty-seven categories (including the LWTs,
28 hybrid types recognised by Lamb (1972)). Similar to Unwin (1980), this study used the original
29 eleven LWTs. Jenkinson's objective daily synoptic indices from 1st January 1880 to 31st July 2007
30 were downloaded from the Climatic Research Unit, University of East Anglia
31 (<http://www.cru.uea.ac.uk/>). These indices were generated based on the dominant airflow direction
32 and vorticity (O'Hare et al., 2005). Considering the data availability of Jenkinson's objective daily
33 synoptic indices till 31st July 2007 and the data availability of the MODIS LST from 8th July 2002,

1 the analysis of this study is performed over the period from 8th July 2002 to 31st July 2007, a total of
2 1850 days. During the study period, the days are associated with the original eleven LWTs account
3 for approximately 73.6% of total days. The remaining hybrid types (anticyclonic-related ANE, AE,
4 ASE, AS, ASW, AW, ANW and AN; cyclonic-related CNE, CE, CSE, CS, CSW, CW, CNW and
5 CN) accounting for 26.4% of the total days are not discussed in this study.

6

7 **2.5 Methods**

8 The UHI intensity (UHII) in this study is defined as the temperature difference between the
9 Edgbaston (urban) station and the Shawbury (rural) station. To examine the characteristics of
10 Birmingham's UHIs (aUHI and sUHI), both ground-based MIDAS T_{air} and satellite-derived MODIS
11 LST were analysed. To define a heat island event, a threshold value was selected for daily $T_{\text{air}/\text{min}}$.
12 Days with nocturnal aUHI intensity (aUHII) in excess of 1.5 °C can be regarded as UHI events
13 (Unwin, 1980). A value of 5 °C was chosen as a threshold for extreme UHI events. Nocturnal aUHII
14 were grouped by LWT and Analysis of Variance (ANOVA) was conducted. The downloaded
15 MODIS Scientific Data Set (SDS) in Hierarchical Data Format (HDF) with sinusoidal grid
16 projections were converted to ArcGIS raster format using the Marine Geospace Ecology Tools
17 (MGET)'s Convert SDS in HDF to ArcGIS Raster tool (Roberts et al., 2010). Once the LST images
18 were converted to ArcInfo Binary Grid format, the spatial patterns were thereafter mapped by
19 combining the outputs with Birmingham's shapefile obtained from EDINA Digimap
20 (<http://digimap.edina.ac.uk/boundary/>). For the LWTs analysis, the MODIS satellite images were
21 selected if they met the following three criteria: (i) complete pixel coverage of nocturnal LST across
22 Birmingham, (ii) clear-sky conditions (0-okta cloud cover), and (iii) no unrealistically extreme
23 high/low LST values. It is obvious that clouds may considerably affect the quality of satellite LST
24 measurements. A total of 77 nocturnal images matched the three criteria. Among the 77 nights, 48
25 nights belong to LWT [A] type, 2 nights to [C], 9 nights to [SW], 2 nights to [W], 4 nights to [NW],
26 3 nights to [S], 2 nights to [N], 3 nights to [SE], 2 nights to [U], 1 night to [E] and 1 night to [NE].
27 To obtain the spatial patterns of Birmingham's sUHI for each LWT, the MODIS LST images for that
28 LWT were averaged by using ArcGIS Map Algebra - Raster Calculator. To obtain the sUHI intensity
29 (sUHII), ArcGIS Zonal Statistics - Spatial Analyst was performed to derive the LST from the urban
30 Edgbaston station and the rural Shawbury station. In addition, the spatial patterns of sUHII were
31 defined by subtracting the LST value retrieved at Shawbury from the 2-D LST.

1 **3. Results and discussion**

2 **3.1 Ground-based aUHI**

3 *3.1.1 Variations of aUHI*

4 Figure 3 represents the monthly variations of aUHII for the period 2002-2007 (N = 1850); error bars
5 refer to the 95% Confidence Interval (CI) based on the standard error estimates of the sample means.
6 The results show consistently stronger nocturnal aUHI intensities than daytime aUHI intensities for
7 all seasons (spring: MAM; summer: JJA; autumn: SON; winter: DJF). In the period from May to
8 August, the daytime aUHI intensities are positive. Statistical analysis of daytime aUHII provides an
9 annual mean value of -0.2 °C. The negative values of daytime aUHII from October to April (or
10 November, January and February, if the elevation correction is made) imply an urban cold island and
11 likely attributed to urban shading. Due to the study core of heat islands, the weak (or absent) daytime
12 aUHI will not be discussed further.

13 Table 1 lists the seasonal mean values and standard deviations of nocturnal aUHII, and the numbers
14 of heat island events together with extreme heat island events over 1850 days. Among four seasons,
15 mean nocturnal aUHII is at a maximum in summer (1.7 °C) and at a minimum in winter (0.8 °C). In
16 view of this, the aUHI is most prominent for summer nights but weakest in winter. This is because
17 heat absorbed during the daytime is highest in summer and lowest in winter. During the study period,
18 there was an average of 39.8% (N = 737) heat island events and 3.0% (N = 58) extreme heat island
19 events. Heat island events most commonly occurred in summer (47.7%), followed by autumn (43.5%)
20 and spring (39.3%), and least commonly occurred in winter with a proportion of 28.2%. Extreme
21 heat island events mainly occurred in spring (3.9%) and summer (3.7%).

22

23 *3.1.2 Nocturnal aUHI by LWT*

24 Unwin (1980) analysed Birmingham's aUHI using Edgbaston as urban station and Elmdon as rural
25 station, and classified the data into LWTs for a period of 1965-1974. In this study, a similar analysis
26 is conducted for the West Midlands using the data detailed in Section 2.4.

27 Table 2 shows the summary of nocturnal aUHII, total numbers of nights and (extreme) UHI events
28 for each LWT. Type [A], [C], [SW] and [W] have been recognised as the most commonly occurring
29 weather types for UHI, having 21.1%, 11.7%, 9.4% and 8.7%, respectively. The results are
30 consistent with the findings of O'Hare et al. (2005) who found that the top three most frequent
31 weather types are [A], [W] and [C] accounting for approximately 50% of total days over the British
32 Isles. Unlike Unwin (1980)'s findings that [W] (22.3%) and [A] (18.4%) were top two most

1 frequently occurring types during 1965-1974, for our study period 2002-2007, [A] covered the
2 highest frequency percentage of 21.1% but [W] only covered 8.7%. Unwin (1980) also found that [C]
3 was associated with least nocturnal UHI effect (0.49 K), our results show that [C] was associated
4 with a mean nocturnal aUHII of 0.9 °C which is higher than [N], [NE] and [E]. The 9.4% in
5 frequency of [SW] is more than doubled compared to Unwin (1980)'s frequency percentage of 3.9%.
6 The most dominant LWT, [A], was associated with a mean nocturnal aUHII of 2.5 °C, the largest
7 proportion of 65.2% of UHI events and 12.0% of extreme UHI events among all LWTs, and the
8 maximum aUHII of 7 °C. Two least dominant LWTs, [NE] and [E], were associated with mean
9 nocturnal aUHII of 0.7 °C and 0.6 °C, respectively. Although the LWT of [U] had low frequency, its
10 mean nocturnal aUHII reached 2.2 °C. During the 30 [U] days, approximately 60% (17 days)
11 occurred in summer. For many other LWTs, [C] is positively skewed peaking at 'no UHI effect'
12 aUHII of 0. The results of one-way ANOVA analysis confirm that the eleven LWTs explain the
13 significant fraction of the variance in aUHI ($p < 0.01$), the excluded hybrid LWTs are statistically
14 insignificant ($p > 0.05$).

15 It should be noted that, if a systematic difference in temperature between the two sites caused by
16 factors other than land use is considered (Peterson and Owen, 2005), for example, the climatological
17 temperature lapse rate due to the difference in elevation, interpretations throughout the section may
18 be slightly altered. If a climatological lapse rate of temperature of $6^{\circ}\text{C km}^{-1}$ is used, aUHII would be
19 0.53°C higher since the elevation of the Edgbaston site is 160 m and that of the Shawbury site is 72
20 m. In other words, the UHII calculated from the direct temperature measurements at the two sites
21 underestimates the true UHII. If this correction were taken, the horizontal line of 'aUHII = 0' in
22 Figure 3 would be moved down to the location of -0.53°C (as indicated by the dashed line) and
23 equivalently all values of aUHII should be increased by 0.53°C .

24

25 **3.2 Satellite-derived sUHI**

26 Figure 4 shows the averaged spatial patterns of Birmingham's sUHII calculated from MODIS/Aqua
27 nocturnal LST for each LWT, indicating Birmingham's urban warming relative to the surrounding
28 rural areas. The number at the lower-right corner of each image is the number of nights to be
29 averaged. A 'concentric pattern' is found in Birmingham central areas with sUHII decreasing
30 towards the sub-urban or rural areas, particularly when influenced by [A], [SW] and [S] types.
31 Birmingham's urban warming of up to 4.16°C has been detected under the influence of [A] type in
32 Fig. 4 [A]. It must be noted that this UHII of 4.16°C is the averaged value of 48 nights and this pixel
33 is not the one covering the Edgbaston station. In Section 3.3 we will show that the averaged sUHII

1 for the pixel of Edgbaston is 3.0 °C, lower than this maximum value of 4.16 °C. Heat islands induced
2 by [SE], [C] and [U] types can also be recognised but at a relatively weak level. Results highlight the
3 significance of [A] type in dominating Birmingham's sUHI. Unfortunately for some LWTs ([C], [W],
4 [N], [U], [NE] and [E]), only a small number of nights satisfied the three criteria and thus
5 interpretation of the results should be made with caution.

6

7 **3.3 Comparison of aUHI and sUHI for cloudless anticyclonic conditions**

8 Because T_{air} is directly measured from a single-point weather station but LST is indirectly measured
9 using satellite remote sensing at the 1 km² pixel spatial resolution, interpretation of the comparison
10 between T_{air} and LST should be made cautiously. In principle, the LST reading of a pixel can be
11 interpreted as an aggregated temperature of all upward-facing surfaces that the satellite sensor 'sees'
12 (e.g. roads, roofs, tree tops, vegetation covers, etc.) inside the pixel. For an urban Edgbaston station
13 shown in Figure 2(a), whereas T_{air} is the air temperature inside the urban canopy layer, the LST
14 obtained from the MODIS/Aqua sensor may be influenced by rapidly cooled exposed surfaces, such
15 as park areas, tree tops and even water bodies (reservoir, lake, etc.).

16 In this section, nocturnal LST and T_{air} at 01:30 UTC of the 48 cloudless anticyclonic nights at both
17 the Edgbaston and Shawbury stations are analysed. In Figure 5, the averaged nocturnal LST and T_{air}
18 for each month are plotted together with the averaged aUHII (01:30 UTC) and sUHII; also plotted as
19 bars are the number of cloudless anticyclonic nights of each month. It is shown that the spring season
20 (Mar-May) has the highest number of cloudless anticyclonic nights, whilst the winter season (Nov-
21 Jan) has the lowest number. The highest sUHI is 5.9 °C occurred on 2 September 2002 and the
22 highest aUHI is 6.8 °C occurred on 11 April 2007 (not shown). Figure 5 also shows that, although
23 LST and T_{air} follow the normal seasonal pattern, UHI intensities do not vary significantly across the
24 seasons. By excluding those months with only one or two nights (Jan, Aug, Oct, Nov, and Dec), we
25 notice a difference of sUHII between summer months and spring months (i.e. higher in summer), but
26 this difference is not as significant for aUHII. Mean aUHII for all 48 nights is 3.4 °C, which is
27 slightly higher than the mean sUHII of 3.0 °C by 0.4 °C.

28 Figure 5 illustrates that for both stations, T_{air} is normally higher than LST (with only one exception
29 of Shawbury in November). For the urban site this may be attributed to the fact that LST is
30 influenced by the cold exposed surfaces of the pixel; for the rural site, however, nighttime radiation
31 inversion could be the explanation of the warmer T_{air} than LST. Figure 6 (a) and (b) demonstrate
32 the correlations between T_{air} and LST of all 48 nights for the two stations, respectively. The
33 intercepts at LST=0 °C are interpreted as the difference between T_{air} and LST if the regression line

1 has the slope of unity: 2.2 °C for the Edgbaston site and 1.6 °C for the Shawbury site. The slope of
2 the regression line for Shawbury is indeed about 1, suggesting that the lower LST due to radiation
3 cooling is independent season. On the other hand, the slope of the regression line for Edgbaston is
4 0.96, suggesting that the difference between T_{air} and LST is weakly season-dependent, reduced from
5 2.2 °C at LST=0 °C to 1.4 °C LST=20 °C. For each station, nocturnal T_{air} and LST are strongly
6 correlated at the 0.01 significance level ($r > 0.98$ for Edgbaston, $r > 0.97$ for Shawbury, $p < 0.01$).

7 The scatter plot of nocturnal aUHII and sUHII for the 48 nights is shown in Figure 6 (c), together
8 with the linear regression results. The correlation is statistically significant at the 0.01 level ($r =$
9 0.419, $p < 0.01$) and accounted for approximately 18% of the variance of aUHII ($R^2 = 0.175$). The
10 linear relationship between sUHI and aUHI can be explained by a simple analytical model as
11 described in Box 1. This model is derived from a few assumptions which are reasonably justifiable.
12 The model is expressed by Equation (8) in Box 1:

$$13 \quad \text{aUHII} = \frac{1}{\alpha} \cdot \text{sUHI}$$

14 which implies that aUHII and sUHII are linearly related with a slope of $1/\alpha$, where $\alpha = \Delta f^{(b)}$ is the
15 difference of “built-up” area fraction between the urban pixel and the rural pixel. Referring to Figure
16 2, $\Delta f^{(b)}$ is positive due to a greater coverage of urban “built-up” landuse type for the Edgbaston
17 pixel than that for the Shawbury pixel. Although this model explains the linear relationship between
18 sUHII and aUHII, the magnitude of the slope is larger than that in Figure 6. We attribute this
19 mismatch to the inaccuracy of the assumptions. One example is the inaccuracy of Equation (6) in
20 Box 1, as we know that under cloudless conditions, normally $T_{\text{air}(r)} > T^{(o)}$ due to radiation
21 inversion (as discussed earlier). This assumption replaces $T^{(o)}$ (a lower value) in Equation (4) by
22 $T_{\text{air}(r)}$ (a larger value) and causes a smaller $(T^{(b)} - T^{(o)})$. To compensate this, the coefficient α in
23 Equation (7) should be larger, consequently leading to a smaller slope of $1/\alpha$. These findings of
24 aUHII-sUHII relationships may lead to a future development of deriving the spatial patterns of aUHI
25 using satellite-derived thermal images.

26

27 **4. Conclusions**

28 This study investigated the relationship between aUHII and sUHII (derived from ground-based T_{air} at
29 two weather stations and satellite-sensed LST, respectively) in relation to Lamb weather types
30 (LWTs) over the period 2002-2007 for Birmingham, UK, with a particular focus on cloudless
31 anticyclonic conditions. Over the study period, the most frequently occurring (391 nights, 21.1% of
32 1850 nights) LWT, ‘anticyclonic’, gives a strongest mean (or maximum) nocturnal aUHII of 2.5 °C

1 (or 7 °C) and the largest proportion of nocturnal heat island events of 65.2% in this LWT. Among the
2 selected 48 cloudless anticyclonic nights, the majority occurred in the spring and summer seasons
3 and both aUHII and sUHII do not vary significantly across the seasons. The averaged sUHII and
4 aUHII over the 48 nights are 3.0 °C and 3.4 °C, respectively. T_{air} is normally higher than LST by
5 about 1.4-2.2 °C. The scatter plot of nocturnal aUHII and sUHII for the 48 nights demonstrates a
6 linear aUHII-sUHII relationship. We also developed a simple analytical model that links the slope of
7 the aUHII-sUHII relationship to the difference of “built-up” area fraction between the urban pixel
8 and the rural pixel. This partially explains the physical basis behind the relationship.

9 In June 2011, two new weather stations were installed: one at the University of Birmingham
10 (Winterbourne 2) and one at Paradise Circus in Birmingham city centre (the Edgbaston weather
11 station has now closed). These two new stations are expected to provide more relevant urban weather
12 data and better information for further research on Birmingham’s UHI and the effect of climate
13 change on extreme weather events. Since the UHI phenomenon describes the local-scale (regional)
14 urban warming, the UHII calculated from different pairs of urban-rural stations could produce
15 different UHI effects. Wan et al. (2002) raised an issue of the impact of rural station selection on
16 UHII estimation. For the future research, more urban and rural weather stations will be considered to
17 compare the UHI effect.

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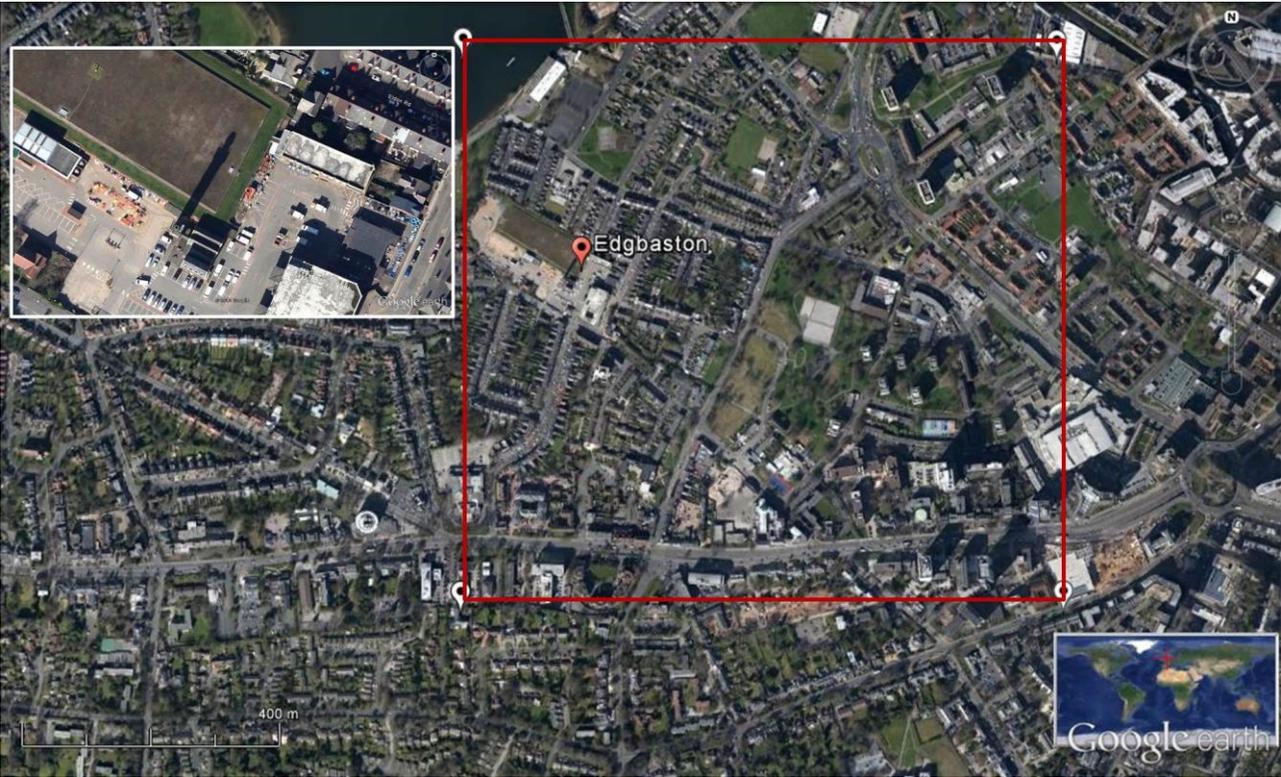
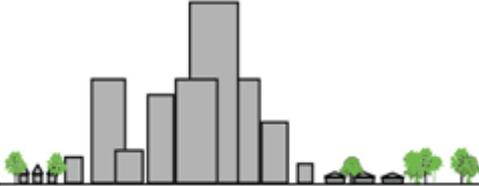
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Figure 7. A simple model in relation to the fraction (%) of area of urban canopy



Figure 1. Map of the West Midlands, UK showing the locations of weather stations (the dark inset boundary surrounding Edgbaston is the West Midlands metropolitan county)

(a) Urban



(b) Rural



Figure 2. Satellite images showing the surroundings of local weather stations: (a) Edgbaston and (b) Shawbury, corresponding to the MODIS 1 km² pixels

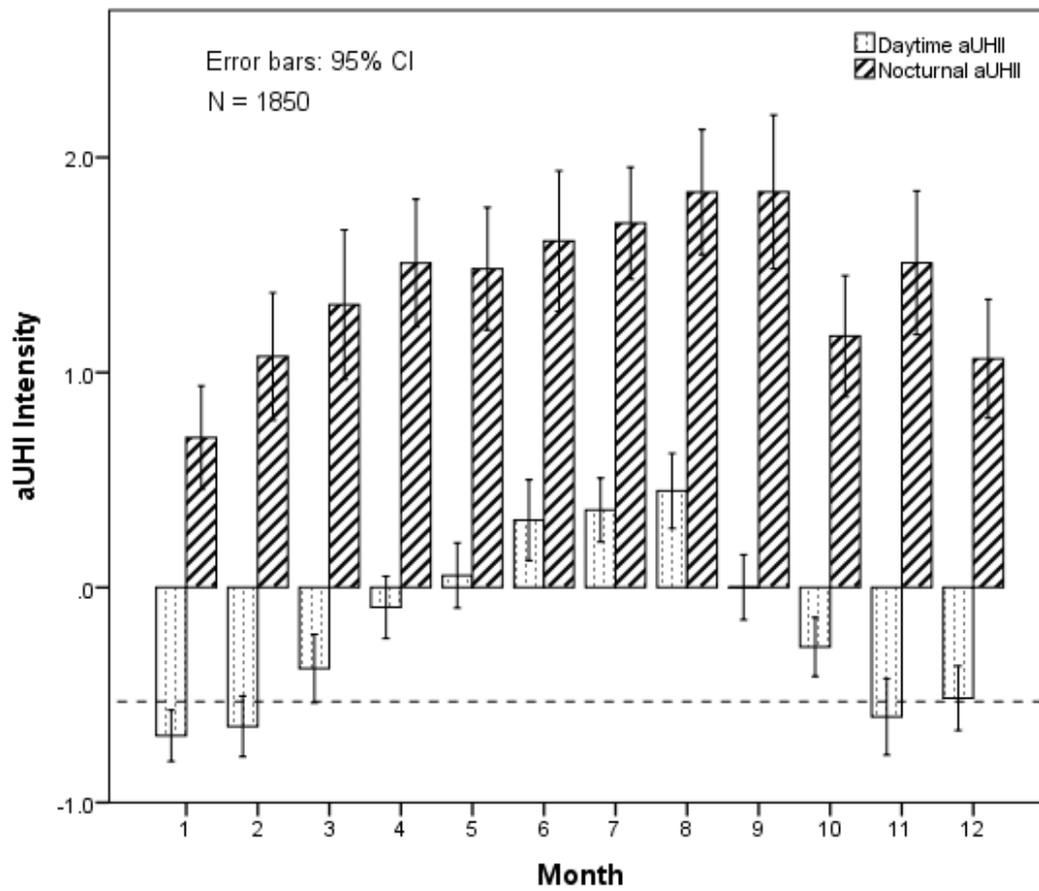
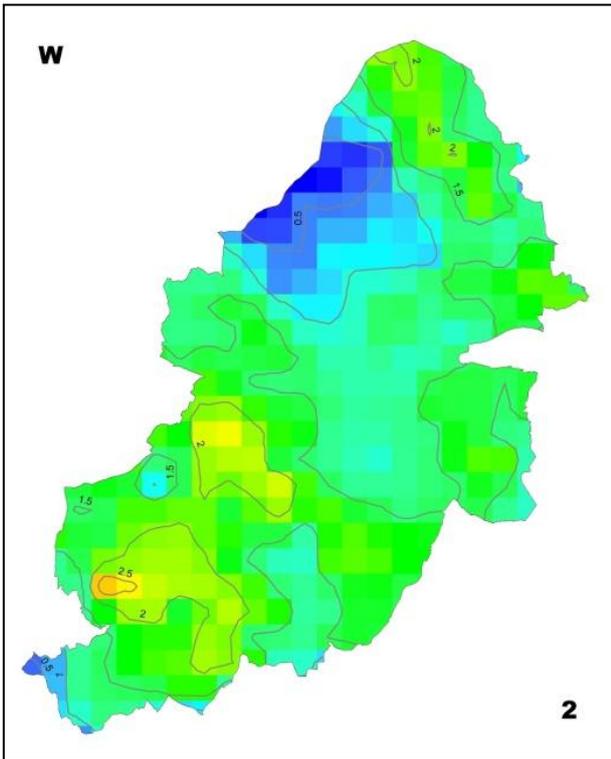
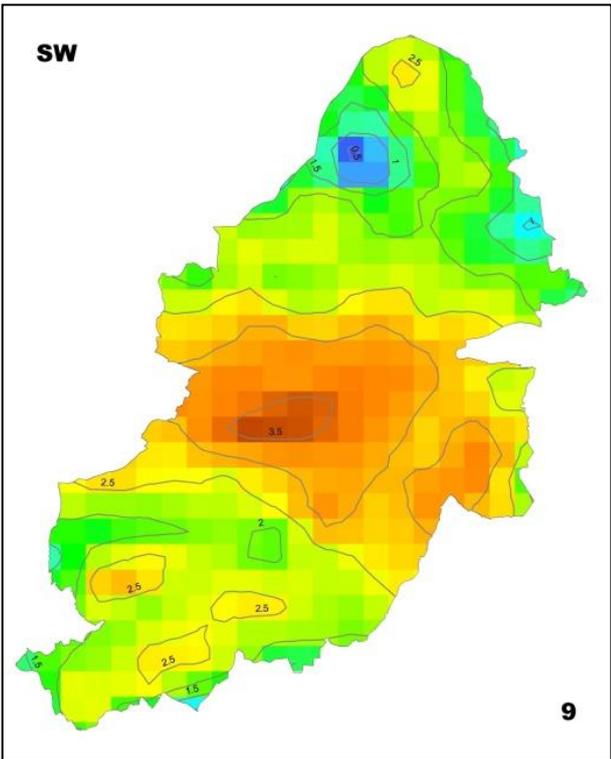
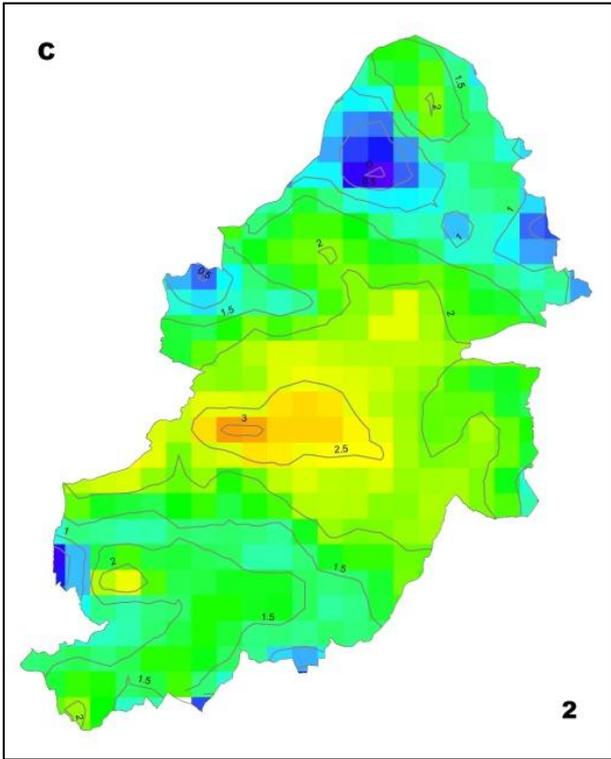
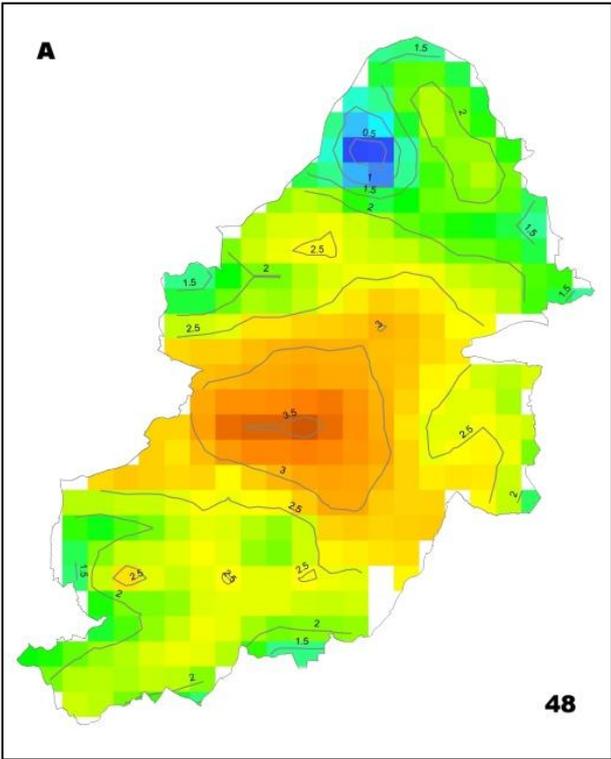
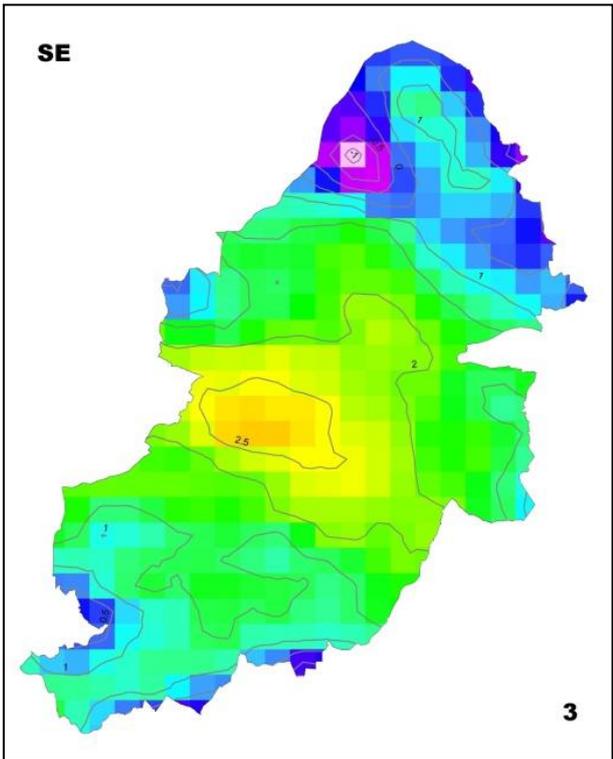
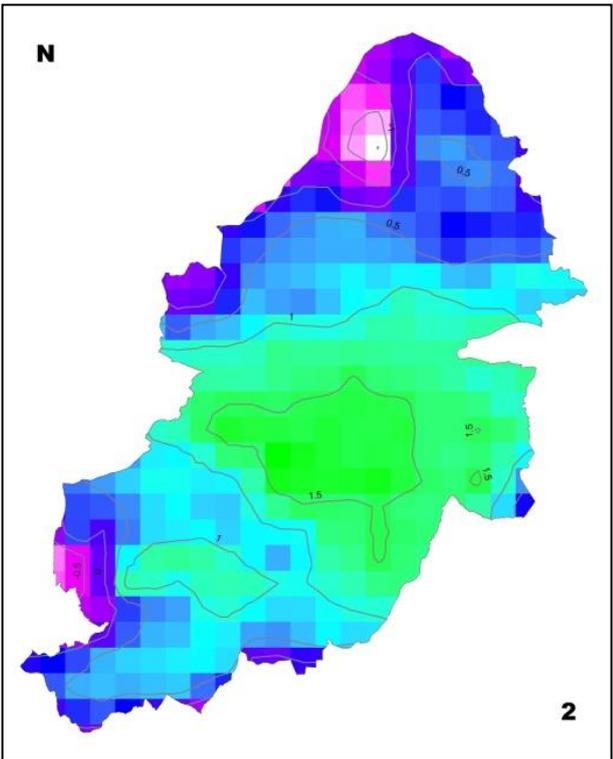
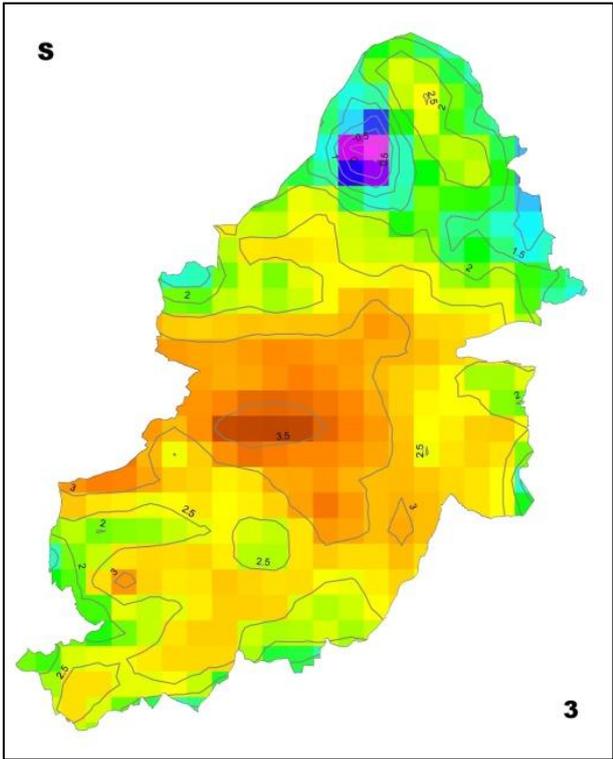
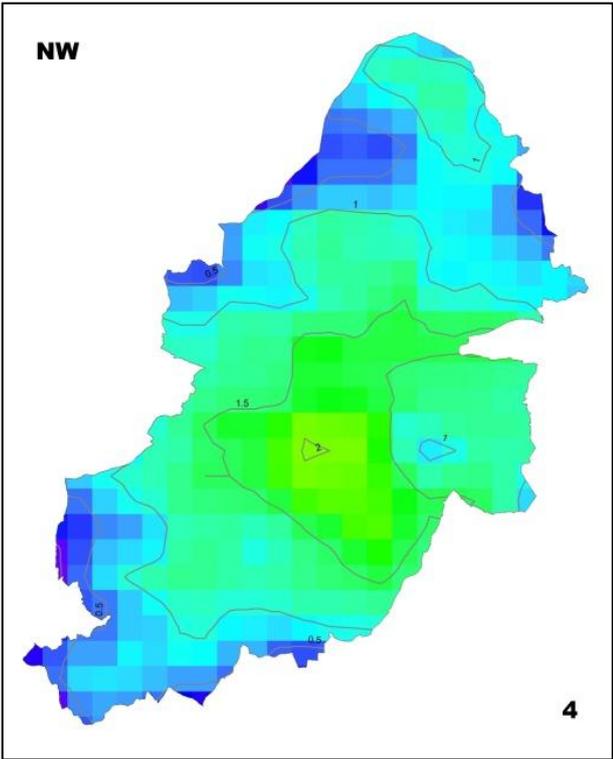


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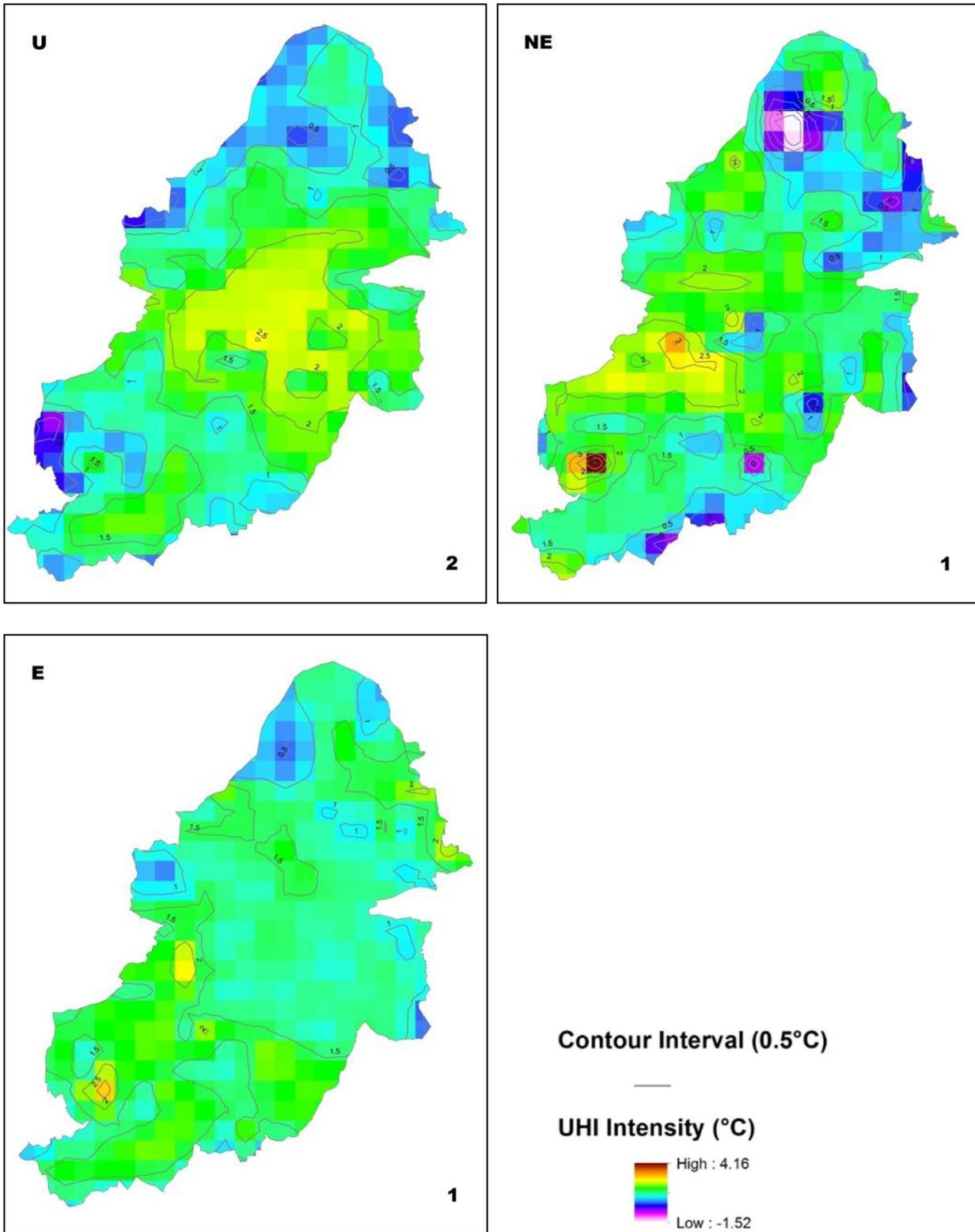


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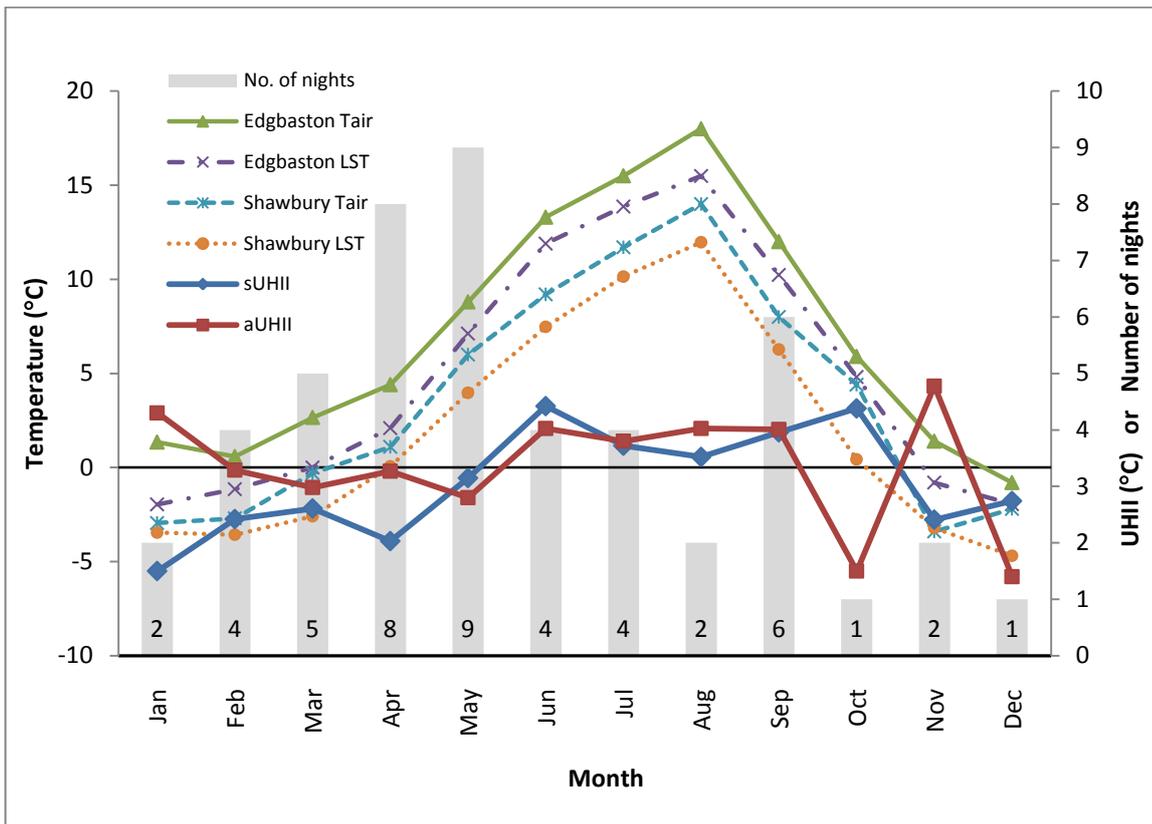
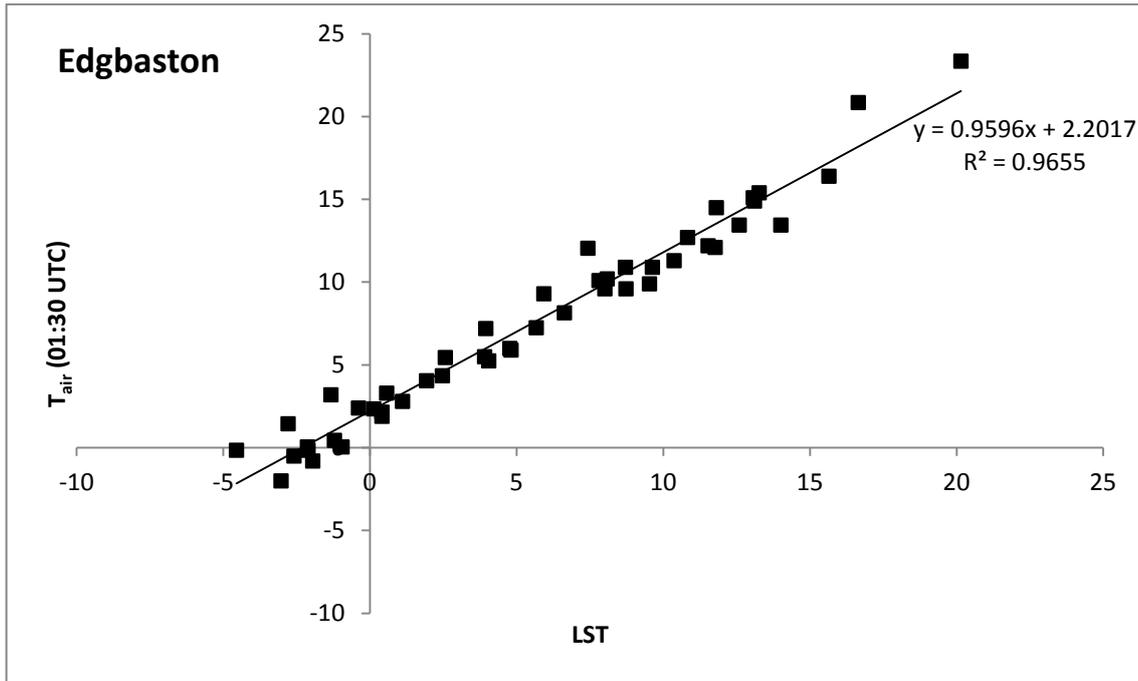
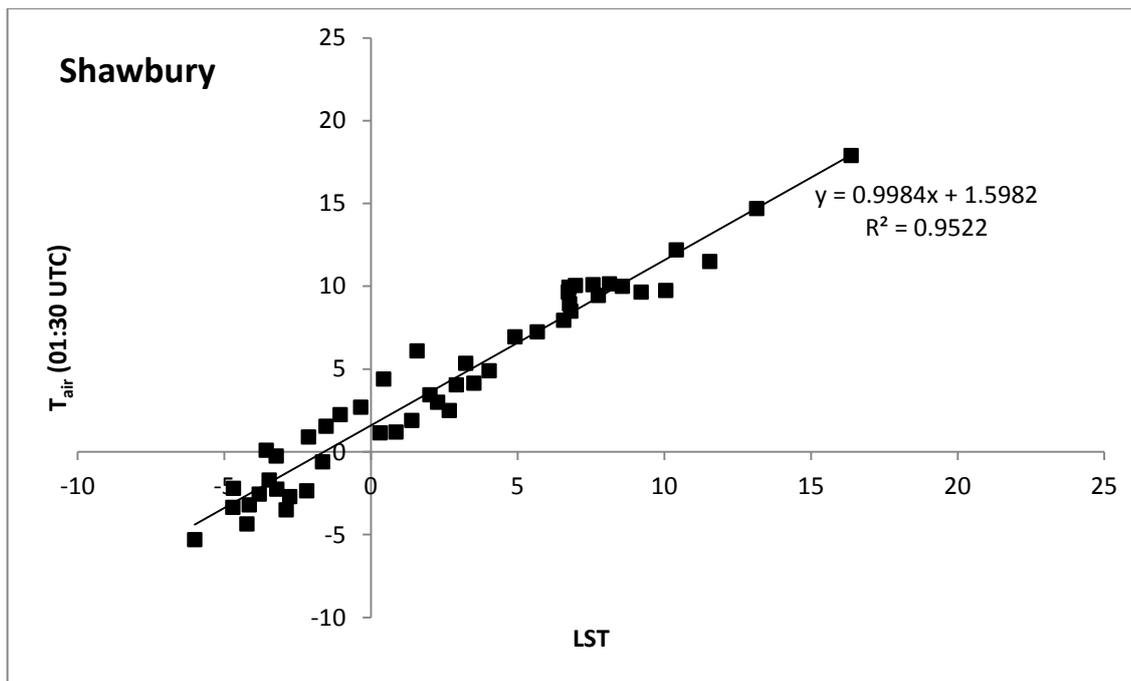


Figure 5. Monthly mean T_{air} (01:30 UTC) and LST at the Edgbaston and Shawbury stations, aUHII and sUHII for the 48 cloudless anticyclonic nights. The bars and the numbers inside indicate the number of nights of each month.

(a)



(b)



(c)

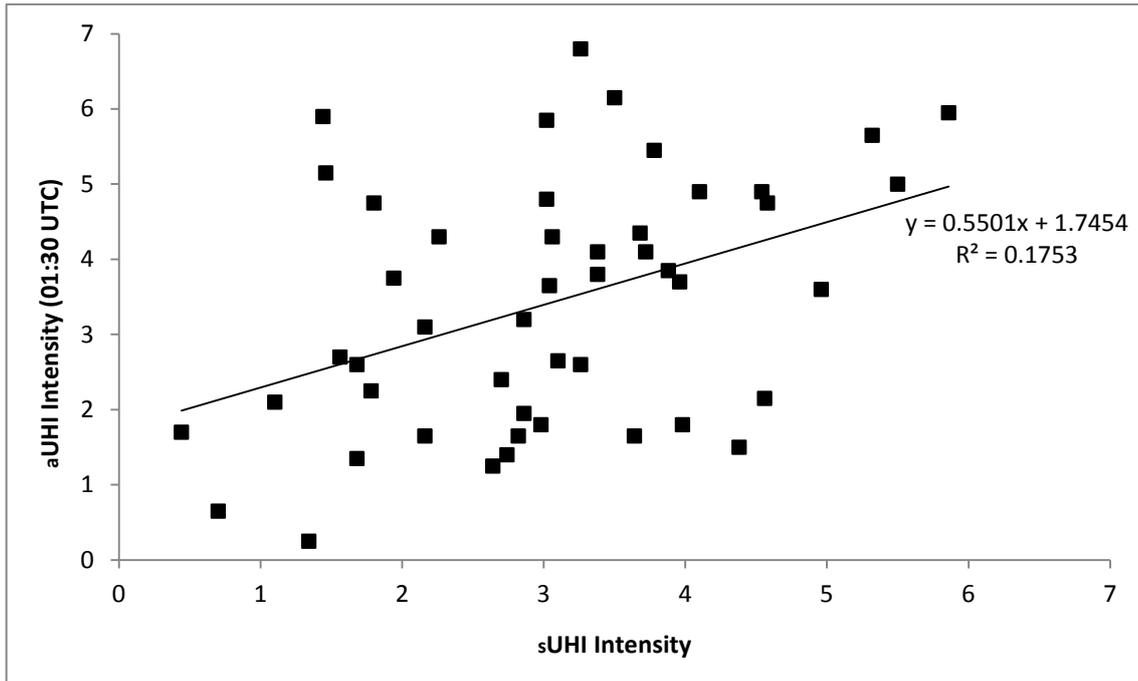


Figure 6. Relationships between nocturnal T_{air} and LST at (a) Edgbaston and (b) Shawbury; (c) Relationship between aUHII and sUHII.

Box 1: Linear relationship between sUHII and aUHII

Assumption 1: We assume that the ground surface of a pixel of the satellite imagery consists of two landuse types: “built-up” (buildings, streets, etc.) and “others” (parks, open ground etc.). We also assume that the satellite-derived land surface temperature of a pixel is a weighted average of two temperatures, one for the “built-up” surfaces, $T^{(b)}$, and the other for the “others” surfaces, $T^{(o)}$:

$$LST = f^{(b)}T^{(b)} + (1 - f^{(b)})T^{(o)} \quad (1)$$

where $f^{(b)}$ is the area fraction (%) of the pixel taken by “built-up” landuse type, the superscript $^{(b)}$ denotes “built-up”, and the superscript $^{(o)}$ denotes “others”. For an urban pixel (e.g. the pixel for the Edgbaston station shown in Fig. 2(a)), Equation (1) becomes

$$LST_{(u)} = f_{(u)}^{(b)}T^{(b)} + (1 - f_{(u)}^{(b)})T^{(o)} \quad (2)$$

and for an rural pixel (e.g. the pixel for the Shawbury station shown in Fig. 2(b)), Equation (1) becomes

$$LST_{(r)} = f_{(r)}^{(b)}T^{(b)} + (1 - f_{(r)}^{(b)})T^{(o)} \quad (3)$$

where the subscript $_{(u)}$ denotes an “urban pixel”, the subscript $_{(r)}$ denotes a “rural pixel”. It is noted that $f_{(r)}^{(b)}$ can be zero if a rural pixel contains no built-up landuse type (i.e. covered entirely by an open ground). Using the definition of $sUHII = LST_{(u)} - LST_{(r)}$, the subtraction of (3) from (2) yields

$$sUHII = \alpha \cdot (T^{(b)} - T^{(o)}) \quad (4)$$

where, $\alpha = \Delta f^{(b)} = f_{(u)}^{(b)} - f_{(r)}^{(b)}$, interpreted as the difference of “built-up” area fraction between the urban pixel and the rural pixel.

Assumption 2: We further assume that at night, the air temperature measured at an urban weather station, $T_{air(u)}$, is in equilibrium with the surface temperature of the “built-up” landuse type, $T^{(b)}$, i.e.,

$$T_{air(u)} = T^{(b)} \quad (5)$$

and that air temperature measured at a rural weather station, $T_{air(r)}$, is in equilibrium with the surface temperature of the “others” landuse type, $T^{(o)}$, i.e.,

$$T_{air(r)} = T^{(o)} \quad (6)$$

Recalling the definition of $aUHII = T_{air(u)} - T_{air(r)}$, substitution of (5) and (6) into (4) yields,

$$sUHII = \alpha \cdot aUHII \quad (7)$$

This equation demonstrates that sUHII and aUHII has a linear relationship with a slope of $\Delta f^{(b)}$. In order to match the relationship in Figure 6, Equation (7) can be rearranged as:

$$aUHII = \frac{1}{\alpha} \cdot sUHII \quad (8)$$

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Table 1. Seasonal mean nocturnal aUHII and (extreme) heat island events, 8th July 2002 to 31st July 2007

Table 2. Summary of nocturnal aUHII, numbers of nights (frequency percentage) and (extreme) heat island events according to LWTs (1850 nights)

Table 1. Seasonal mean nocturnal aUHII and (extreme) heat island events, 8th July 2002 to 31st July 2007

Season	Nocturnal aUHII	A. No. nights	B. No. UHI nights >1.5 °C (% of column A)	C. No. extreme UHI nights >5.0 °C (% of column A)
Spring (MAM)	1.4 (1.80)	460	181 (39.3%)	18 (3.9%)
Summer (JJA)	1.7 (1.63)	484	231 (47.7%)	18 (3.7%)
Autumn (SON)	1.5 (1.82)	455	198 (43.5%)	14 (3.1%)
Winter (DJF)	0.8 (1.61)	451	127 (28.2%)	8 (1.8%)
Total	1.4 (1.91)	1850	737 (39.8%)	58 (3.0%)

(Bracketed values are standard deviations of aUHII)

Table 2. Summary of nocturnal aUHI, number of nights (frequency percentage) and (extreme) heat island events according to LWTs (1850 nights)

Lamb weather types	Nocturnal aUHI				A.	B.	C.
	Min	Max	Mean	Std. Deviation	No. nights (% of 1850 nights)	No. UHI nights >1.5 °C (% of column A)	No. extreme UHI nights >5.0 °C (% of column A)
Anticyclonic [A]	-8.5	7.0	2.5	2.11	391 (21.1%)	255 (65.2%)	47 (12%)
Cyclonic [C]	-1.7	7.3	0.9	1.52	217 (11.7%)	57 (26.3%)	5 (2.3%)
South-westerly [SW]	-2.0	6.7	1.1	1.80	173 (9.4%)	58 (33.5%)	5 (2.9%)
Westerly [W]	-3.3	5.3	1.0	1.66	160 (8.7%)	55 (34.4%)	2 (1.3%)
North-westerly [NW]	-1.7	5.4	1.2	1.40	107 (5.8%)	36 (33.6%)	1 (0.9%)
Southerly [S]	-7.2	7.2	1.1	2.08	104 (5.6%)	32 (30.8%)	4 (3.8%)
Northerly [N]	-1.3	4.9	0.9	1.41	74 (4.0%)	21 (28.4%)	-
South-easterly [SE]	-0.9	5.2	0.9	1.67	54 (2.9%)	14 (25.9%)	1 (1.9%)
Unclassified [U]	-0.9	6.2	2.2	1.93	30 (1.6%)	16 (53.3%)	3 (10%)
North-easterly [NE]	-2.0	4.9	0.7	1.65	25 (1.4%)	5 (20.0%)	-
Easterly [E]	-1.4	4.5	0.6	1.53	25 (1.4%)	5 (20.0%)	-
[A]/[C] hybrids	-11.2	6.5	1.3	1.86	490 (26.4%)	200 (40.8%)	10 (2.0%)