

WUDAPT

An Urban Weather, Climate, and Environmental Modeling Infrastructure for the Anthropocene

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WUDAPT is an international community-generated urban canopy information and modeling infrastructure to facilitate urban-focused climate, weather, air quality, and energy-use modeling application studies

The Anthropocene epoch, the human-influenced geologic time period (Crutzen and Stoermer 2000), is linked inextricably to urbanization. Human activities in this epoch have had a demonstrable impact on climates at all scales, and without proper management, increased urbanization will contribute to associated extreme and unexpected weather events in cities. Currently, more than half of the planet's population resides in urban areas, and by 2050, up to 75% are projected to live in cities of varying sizes (United Nations 2014). The development of ever more powerful computer models to simulate weather and climate, air quality, hydrology, and other environmental processes now allow us to evaluate the impacts of urban areas on climate processes and to assess urban vulnerabilities to natural hazards. These tools are needed to support urban management, to mitigate deleterious effects, and to support resiliency strategies but require climate relevant information on urban landscapes to be effective (Masson et al. 2014).

The effect of urbanization on the environment is an outcome of its physical form (i.e., the land cover, the materials, and the geometry of buildings) and

its functions (the transportation, energy usage, and generation of waste products) that sustain human activities. These vary spatially and temporally and act in concert to adversely affect local climate, hydrology, biodiversity, and air quality. These impact the quality of life and sometimes enhance risks to public health; for example, the urban heat island (UHI) is exacerbated during heat wave events and makes city dwellers especially exposed to heat stress. It is therefore crucial to characterize as best as possible these urban properties, so to be able to predict, via modeling (Chen et al. 2011), the hazard, exposure, and vulnerabilities of urban dwellers to present and future environmental states (NRC 2012). Sustained research on urban meteorology and climate over the past 50 years has provided insights into the layering of the urban boundary layer and its links with the underlying surface (Fig. 1a). As a result, state-of-the-science numerical models can simulate the surface energy budgets, weather, climate, and air quality.

Examples include the Surface Urban Energy and Water Balance Scheme (SUEWS), the Weather Research and Forecasting (WRF) Model, the

Community Earth System Model (CESM), and the Community Multiscale Air Quality (CMAQ) model; each of these systems continue to evolve, providing enhanced capabilities, results, and guidance at increasingly finer grid resolutions. However, these models are reliant on appropriate data that capture the spatially varying and temporally evolving characteristics of urban surfaces; Fig. 1b shows common urban canopy parameters (UCPs) that are needed by “urbanized” climate models. In North America, the National Urban Database and Access Portal Tool (NUDAPT) compiled this information for parts of more than 40 cities (Ching et al. 2009), but in most places the data to derive UCPs are either not available/incomplete and/or available at poor spatial/temporal resolutions. The absence of internationally consistent urban data for such purposes is recognized by global-to-urban climate science communities to be a significant impediment to scientific progress (Jackson et al. 2010; Revi et al. 2014; Baklanov et al. 2009, 2018). Overcoming this impediment is the aim of the World Urban Database and Access Portal Tool (WUDAPT) project.

In this paper, we review the concepts and operational methodologies that underpin WUDAPT (Ching 2013; Ching et al. 2016, 2017b), present some initial results, and present near-term plans. Our intent is to introduce the project and demonstrate its value to the climate community and, while individual experiments are introduced, the research details are referenced rather than discussed in detail.

WUDAPT OVERVIEW. The goals of WUDAPT are 1) to acquire and make accessible coherent and consistent descriptions and information on form and function of urban morphology relevant to climate,

weather, and environment studies on a worldwide basis and 2) to provide a portal with tools that extract relevant urban parameters and properties for models and for model applications at appropriate scales for various climate, weather, environment, and urban planning purposes. Its guiding principle is to generate “fit for purpose” urban data using a globally consistent methodology using available, publicly accessible input data and tools. Products created from this process are shared across multiple communities and platforms.

The data needed to apply models successfully to cities must meet several criteria. First, the modeling description of the urban surface must permit the model to resolve the temporal and spatial characteristics of the mesoscale urban boundary layer, including properties at local scales (Fig. 1a). Second, the spatial gradients of the input (and thus the output) fields are typically highly variable across urban landscapes; consequently, any coarse model grid must represent subgrid variations (Ching 2013; Mouzourides et al. 2013, 2014). Third, data requirements for urbanized models can be highly specialized; typically, they are distinguished by their need for UCP information on building height, vegetative cover, building materials, etc. (see Table 1 and Fig. 1b; Masson 2000; Martilli et al. 2002; Dupont et al. 2004; Otte et al. 2004; Oleson et al. 2008). Fourth, for worldwide applicability, UCPs should be collected using a scheme that is consistent and reliable. Finally, given the time frame, the generation of this database should be practicable and achievable on a reasonably short time frame for greatest impact. WUDAPT adopts a pragmatic approach to meet these criteria.

The components of the urban landscape that are relevant to climate can be organized by scale into

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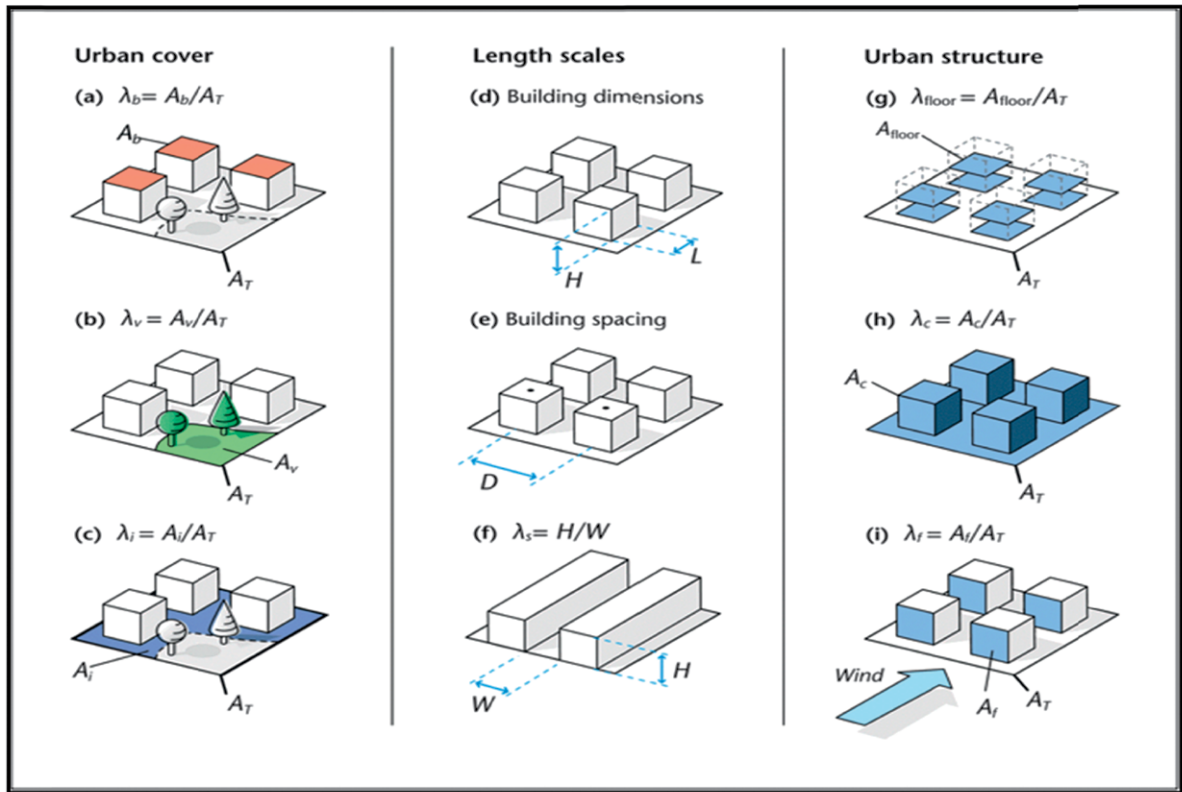
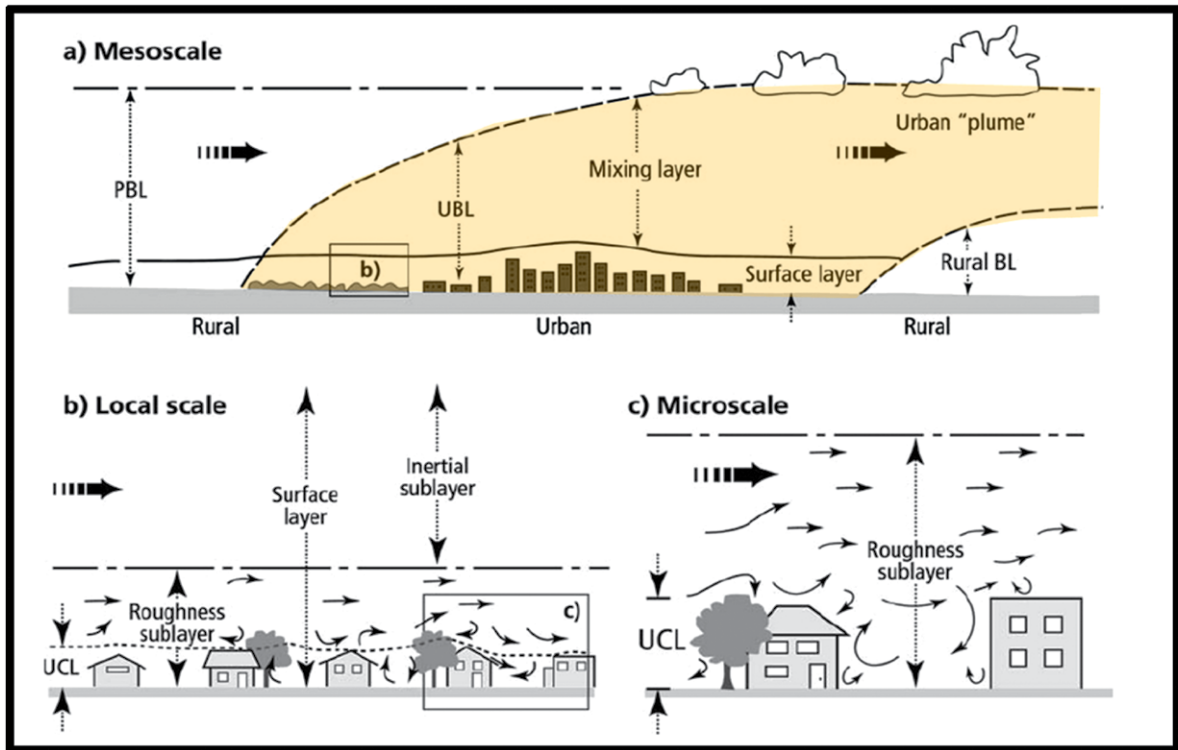


FIG. 1. (top) Structure of the urban boundary layers (source: Oke 2006) showing the developing of the mixed layer above the underlying surface layer in terms of (a) mesoscale, (b) local scale, and (c) microscale, where exchanges are modulated by urban form and functions. (bottom) Common UCPs that describe the character of the urban surface and are employed in models to evaluate the urban effect on wind, temperature, runoff, etc. (courtesy Andreas Christen).

TABLE 1. Examples of UCPs used in urban models. The BEP scheme (Martilli et al. 2002) that is linked to the WRF in W2W specifically utilizes the building UCPs in column 2.

Urban Canopy Parameters		
General	Buildings	Vegetation
Mean canopy height	Mean height	Vegetation plan area density
Canopy plan area density	Std dev of heights	Vegetation top area density
Canopy top area density	Height histogram	Vegetation frontal area density
Canopy frontal area density	Wall-to-plan area ratio	
Roughness length	Height-to-width ratio	Mean orientation of streets
Displacement height	Plan area density	Plan area fraction surface covers
Sky view factor	Rooftop area density	Percent connected impervious areas
	Frontal area density	Building material fraction

facets, elements, streets and blocks, and neighborhoods (Oke et al. 2017). Facets describe flat and uniform features that are distinguished by their slope and aspect and radiative and thermal properties. Elements are the combination of facets that create 3D features like building typologies. Streets and blocks represent the organization of elements to form distinct geometries, and neighborhoods describe a common and repeated amalgam of facets, elements, and streets and blocks over an area. To cope with this complexity, WUDAPT information is organized by level of detail (*L*), and data at each

level are gathered using distinct methodologies and techniques.

The lowest level of detail (L0) maps cities and their surrounding natural landscape into local climate zone (LCZ) types (Stewart and Oke 2012). L1 data use the LCZ maps to provide a sampling context for acquiring and managing information at finer scales. L2 data are complete information on all urban elements (e.g., building footprints, envelope fabrics, and heights) that may exist for some urban areas, albeit coverage is limited to a few cities; for example, NUDAPT data for Houston include detailed information (dimensions

TABLE 2. Some of the UCP values associated with LCZ types from Stewart and Oke (2012). Columns represent the percentage of impervious (λ_i), built (λ_b), and vegetated (λ_v) land cover and mean height of building elements (z), sky view factor (λ_s ; see Fig. 1a), albedo (α), and anthropogenic heat flux (Q_f).

LCZ	λ_i	λ_b	λ_v	z (m)	λ_s	α	Q_f ($W m^{-2}$)
1. Compact high rise	40–60	40–60	<10	>25	0.2–0.4	0.10–0.20	50–300
2. Compact midrise	40–70	30–50	<20	10–25	0.3–0.6	0.10–0.20	<75
3. Compact low rise	40–70	20–50	<30	3–10	0.2–0.6	0.10–0.20	<75
4. Open high rise	20–40	30–40	30–40	>25	0.5–0.7	0.12–0.25	<50
5. Open midrise	20–40	30–50	20–40	10–25	0.5–0.8	0.12–0.25	<25
6. Open low rise	20–40	20–50	30–60	3–10	0.6–0.9	0.12–0.25	<25
7. Lightweight low rise	60–90	<20	<30	2–4	0.2–0.5	0.15–0.35	<35
8. Large low rise	30–50	40–50	<20	3–10	>0.7	0.15–0.25	<50
9. Sparsely built	10–20	<20	60–80	3–10	>0.8	0.12–0.25	<10
10. Heavy industry	20–30	20–40	40–50	5–15	0.6–0.9	0.12–0.20	>300
A. Dense trees	<10	<10	>90	3–30	<0.4	0.10–0.20	0
B. Scattered trees	<10	<10	>90	3–15	0.5–0.8	0.15–0.25	0
C. Bush, scrub	<10	<10	>90	<2	0.7–0.9	0.15–0.30	0
D. Low plants	<10	<10	>90	<1	0.2–0.4	0.15–0.25	0
E. Bare rock or paved	<10	>90	<10	<0.25	>0.9	0.15–0.30	0
F. Bare soil or sand	<10	<10	>90	<0.25	>0.9	0.20–0.35	0
G. Water	<10	<10	>90	—	>0.9	0.02–0.10	0

and construction materials) for every building in the city center (Ching et al. 2009), and Applied Modeling and Urban Planning Laws: Urban Climate and Energy (MApUCE) data comprise a complete inventory of buildings in France (Masson et al. 2015).

The protocols for deriving and using L0 data are now well developed (Bechtel and Daneke 2012; Bechtel et al. 2015, 2017a,b), and there are currently over 80 cities globally for which data are available. The methods for acquiring, managing, and using higher-level data within the WUDAPT framework is being developed (see “Current status and next steps” section), but WUDAPT is already recognized as a framework for urban climate research to integrate more complex physical processes in urban canopy models (e.g., Wouters et al. 2016).

Level 0 data. The Stewart and Oke (2012) LCZ typology was designed primarily to describe the features that impact the near-surface local thermal environment, specifically the roles of land cover and anthropogenic heat on the magnitude of the

observed UHI (e.g., Alexander and Mills 2014). Its outstanding merit is that it is designed as a culturally neutral description of urban landscapes and, critically, each of the 17 basic types (10 of which are urban) is associated with typical value ranges for a set of key urban canopy parameters (Table 2). L0 data are derived using Landsat data, image software, and the knowledge of urban experts (see Bechtel and Daneke 2012; Bechtel et al. 2015, 2017a). The urban experts are critical to the process, as they create the training areas (TAs) that identify the parts of the city under study that exemplify each LCZ type. This information is used to classify Landsat scenes into LCZ maps using a random forest (RF) classifier implemented in the System for Automated Geoscientific Analyses (SAGA) software (Conrad et al. 2015).

The quality of the L0 data relies on the skill of the experts that create the TAs, and considerable effort has been placed on training experts and on independent assessment of the TA data. The current quality control scheme emphasizes the statistical reliability of a city database by randomly dividing

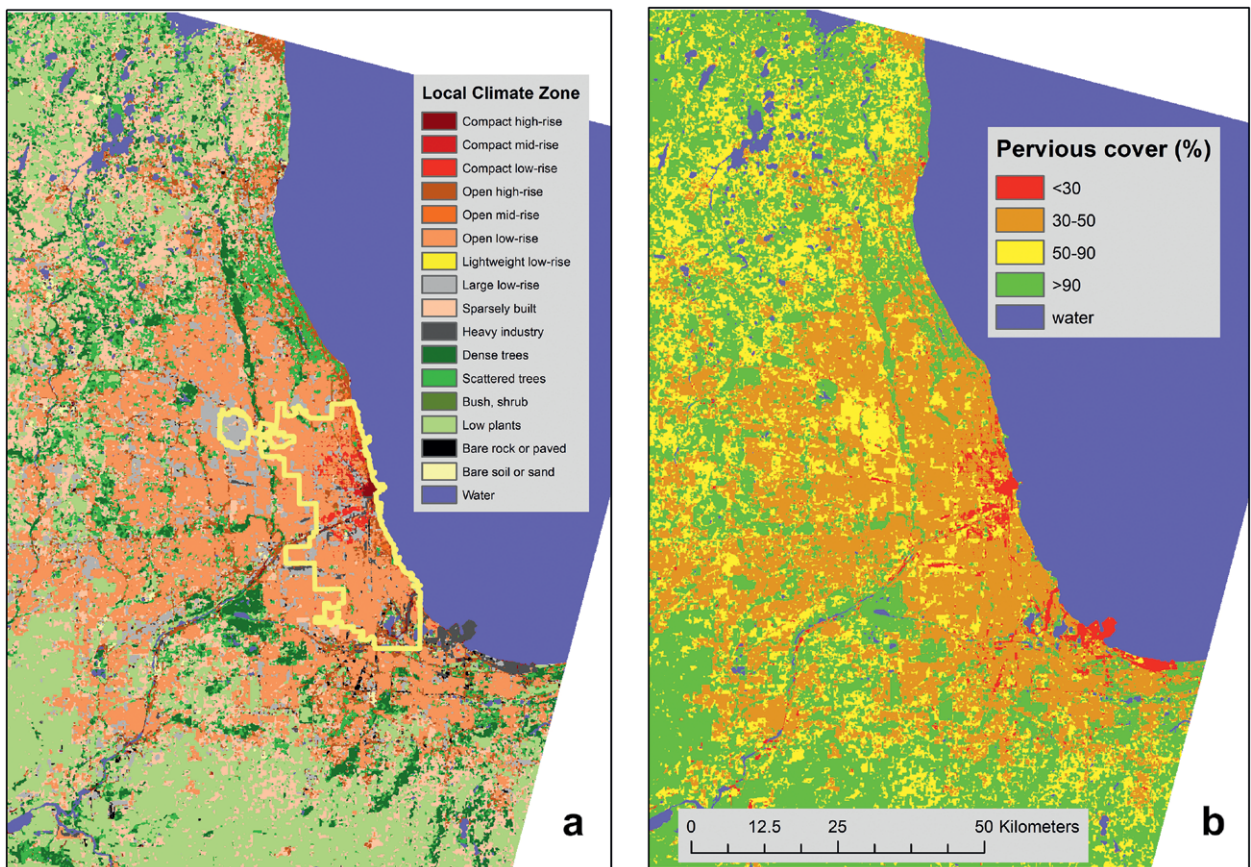


FIG. 2. WUDAPT L0 data for Chicago and derived urban canopy parameters: (a) the distribution and legend of LCZs derived from Landsat images using the WUDAPT protocol and (b) the area fraction of pervious surface cover derived from Table 2.

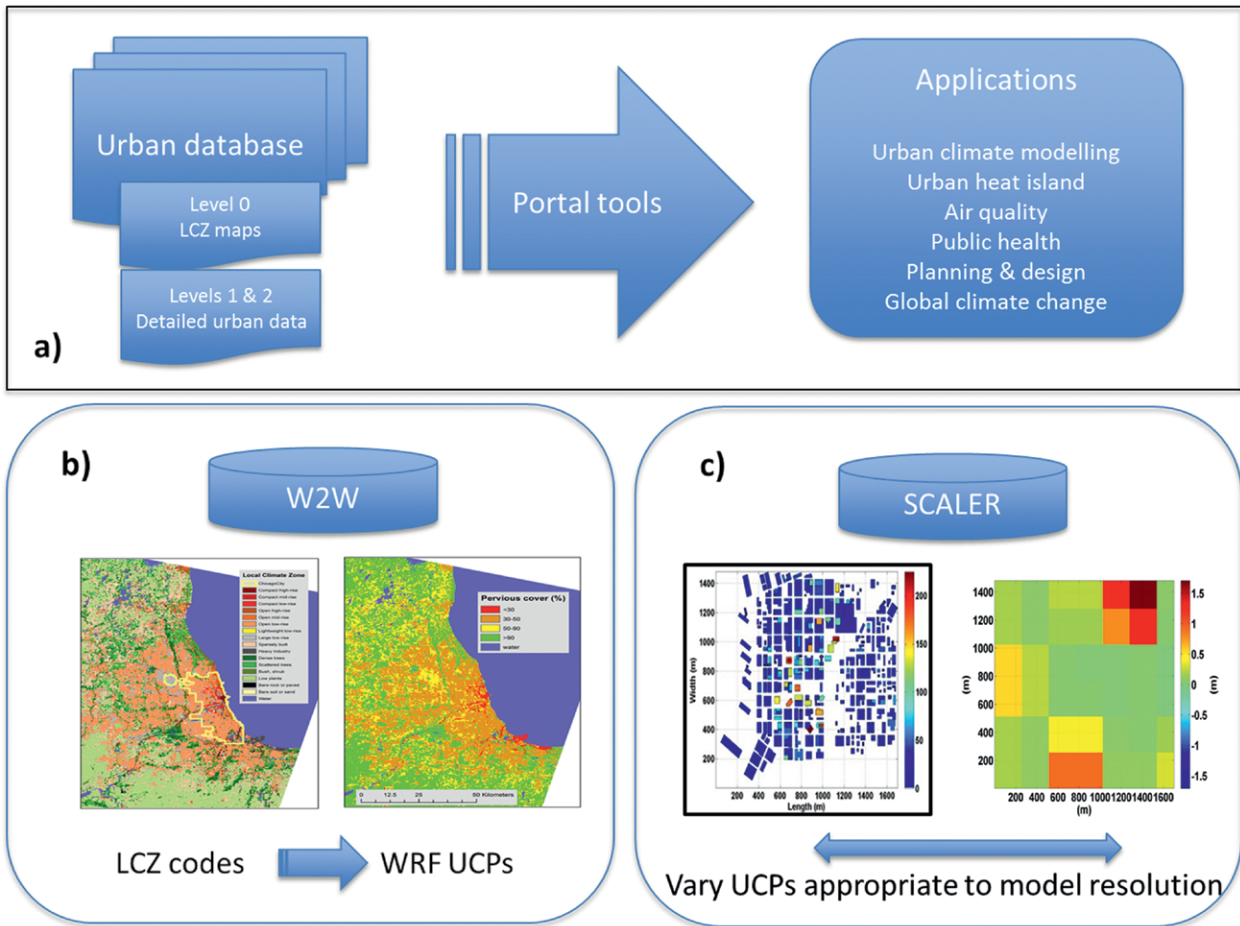


FIG. 3. (a) The schematic of the WUDAPT project and its current portal tools: (b) W2W tool is designed to integrate WUDAPT table lookup UCP data for each LCZ to facilitate its use in the WRF Model and (c) SCALER, a tool that permits the extraction of WUDAPT-type data to user-specified grid resolution.

the TAs into a set for training and a set for evaluation purposes. With each iteration, an LCZ map is generated for a given TA set, and the resulting LCZs are compared with the evaluation set; overall accuracy (OA) is measured as the percent of LCZ values that are predicted correctly. Repeatedly sampling (i.e., bootstrapping) from the TAs allows us to measure the robustness of the LCZ map, that is, the consistency of the LCZ map when using different sets of training areas. A WUDAPT committee that oversees the quality of the L0 data examines the final LCZ map to ensure that it provides an accurate depiction of the urban landscape. There are currently more than 80 cities that are in the WUDAPT database; the reader should refer to the website (www.wudapt.org) for updates.

Each LCZ map encodes UCP values that can be used in models [a subset of the list of parameters is shown in Table 2 and in Stewart and Oke (2012) and its supplemental material]; these UCPs are used in models and climate analyses. As an example, Fig. 2

shows the LCZ map for the Chicago, Illinois, area alongside a map of the pervious fraction that has been generated from a lookup table (Table 2); note that LCZ types are associated with ranges of UCP values. Establishing the veracity of the derived data is not straightforward, as it requires independently derived information that is comparable in scope and spatial resolution. Experiments on a few cities have shown good agreement, but these tests are, to this point, limited to plan area fractions in western cities (Mills et al. 2015, 2017a,b).

The WUDAPT portal. The portal is designed to support climate research that requires urban information (Ching et al. 2015). Critically, it should allow users to extract relevant data at an appropriate spatial scale for modeling purposes. Currently, WUDAPT provides tools that can utilize the L0 data (Fig. 3a), but other tools that require L1/L2 data are being designed; here we describe two portal tools, WUDAPT to WRF (W2W; Fig. 3b) and SCALER (Fig. 3c).

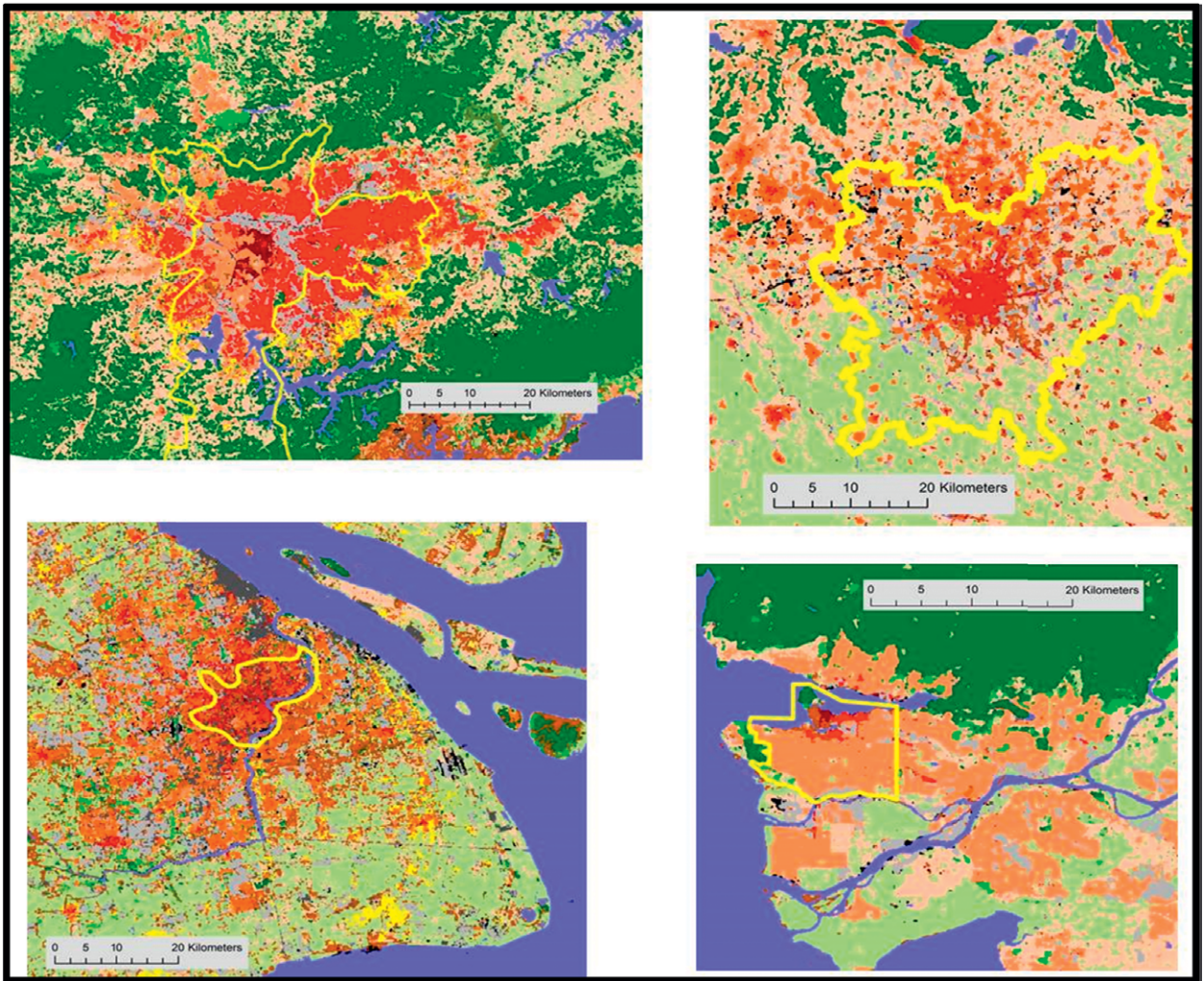


FIG. 4. A comparison of WUDAPT L0 maps for selected cities: (top left) São Paulo, Brazil; (top right) Milan, Italy; (bottom left) Shanghai, China; and (bottom right) Vancouver, Canada. In each case the administrative boundaries of the city or municipality are shown. The LCZ legend is as in Fig. 2a.

The W2W tool was developed to convert L0 data into a gridded format suitable for urban schemes used in the WRF Model; these include the single-layer urban canopy model (Kusaka et al. 2001; Kusaka and Kimura 2004) and the Building Effect Parameterization and Building Energy Model (BEP-BEM) scheme (Martilli et al. 2002; Salamanca et al. 2010). Converting the LCZ parameter information into UCPs suitable for these schemes requires some modification. For example, BEP-BEM requires information on street width, building footprints, and pervious surface cover that can be estimated from the LCZ data by selecting the midpoint values of the available ranges (Table 2). It also requires information on the distribution of building heights within a grid cell, for which there is not a unique solution. The simplest option, which is in use, is to choose three heights, one close to the midpoint value (considering the constraint that it must be a multiple of 5 m)

with a probability of 50% and two other heights above and below that, but within the given range and a multiple of 5 m, with a probability of 25%. The important point, however, is that W2W provides a standardized means for incorporating UCPs into urbanized WRF and permits greater comparability between studies (Brousse et al. 2016); some examples are shown in the next section. Current and subsequent updates of W2W documentation (Martilli et al. 2016) are provided as a link under “Resources” on the WUDAPT website (www.wudapt.org).

SCALER generates appropriately scaled model inputs to various modeling systems (Fig. 3c). This tool uses the principle of the multiple resolution analysis (MRA) to manage the multiscale grid requirements of users (Mouzourides et al. 2013, 2014). Its unique feature is its ability to retain subgrid data on the input parameters as the selected model grid

TABLE 3. The proportion of land area occupied by each LCZ in selected cities. Each city has two values representing the areal percentage in the ROI (left column) and the municipal area (as shown by the yellow boundaries in Fig. 4) for each LCZ. The total area is given in the bottom row.

LCZ type	Chicago		Milan		Shanghai		São Paulo		Vancouver	
Compact high rise	0.20	2.08	0.00	0.00	0.25	4.47	0.51	2.81	0.15	1.36
Compact midrise	0.10	2.39	2.11	6.32	1.35	21.56	0.10	0.31	0.00	0.06
Compact low rise	0.19	4.00	0.10	0.25	0.77	6.23	6.39	26.04	0.47	3.93
Open high rise	3.31	8.39	1.98	5.47	8.66	15.51	1.85	1.16	1.00	4.97
Open midrise	0.14	2.43	7.84	16.13	6.31	15.91	1.01	1.70	0.04	0.03
Open low rise	14.54	53.92	3.57	0.38	2.44	3.20	8.03	15.57	19.97	60.18
Lightweight low rise	0.00	0.00	0.00	0.00	4.35	2.06	1.45	3.45	0.00	0.00
Large low rise	3.58	13.15	4.75	10.70	8.65	5.18	3.63	9.83	3.86	4.82
Sparsely built	13.01	2.78	29.69	33.47	6.16	0.05	22.12	12.54	11.30	3.44
Heavy industry	0.53	3.22	0.00	0.00	3.92	17.10	0.66	0.89	0.00	0.00
Dense trees	3.92	0.93	20.03	0.42	1.19	0.38	39.05	18.11	25.59	1.92
Scattered trees	6.85	2.28	0.47	0.51	1.63	0.31	3.07	1.62	2.02	1.73
Bush, scrub	0.00	0.00	0.00	0.00	0.00	0.00	1.51	0.50	0.00	0.00
Low plants	20.55	1.68	25.34	23.58	26.31	0.23	5.61	1.27	14.36	2.79
Bare rock or paved	0.26	0.80	1.10	2.16	0.76	1.43	0.07	0.05	0.24	0.22
Bare soil or sand	0.54	0.40	0.00	0.00	0.00	0.00	0.44	0.19	0.00	0.00
Water	32.26	1.55	3.02	0.60	27.24	6.37	4.49	3.97	21.01	14.56
Area (km ²)	15,584	597	6,236	1,344	8,887	197	9,278	1,410	2,277	136

scale is increased. This allows the impact of subgrid UCP variability on resulting model outputs to be examined and enables a clearer understanding of the role and impact of such parameters on the behavior of a complex urban system. It has already been used to explore the scale-dependent links between energy demand and urban weather (Neophytou et al. 2015; Mouzourides et al. 2017).

INITIAL ANALYSES AND SAMPLE APPLICATIONS. The innovation of the LCZ scheme explained earlier is that it provides a common platform for comparing cities in terms of urban form and, to a lesser extent, urban function (Stewart and Oke 2012; Gál et al. 2015). Figure 4 shows a sample of LCZ (and their corresponding urban canopy parameters) maps for a variety of cities, revealing their unique and distinct spatial patterns of distribution. Thus, each urban area will have its own unique spatial distribution of urban canopy parameters and, therefore, mesoscale modeling outcomes. The areal coverage for each LCZ type present is shown in Table 3 for both the region of interest (ROI) and official urban administrative area (shown in Fig. 4). Generally, relatively small proportions are occupied by compact urban neighborhoods—the exception is Shanghai, but

it has the smallest area within the official city boundary. Chicago and Vancouver are distinguished by the extent of the open low rise (LCZ 6) and the extent of nearby water. Low plant (LCZ D) characterizes the natural cover outside most cities, but in the case of São Paulo it is dense trees (LCZ A).

These different LCZ geographies should give rise to different urban climate effects. To illustrate, Fig. 5 shows the LCZ maps for São Paulo and Mumbai (India) alongside Moderate Resolution Imaging Spectroradiometer (MODIS)-derived mean annual surface temperature (MAST), which was computed from a 12-yr time series of MODIS land surface temperature acquired at 2230 local time and is a cloud-free, robust, and representative measure of long-term land surface temperature (Bechtel 2015). The spatial pattern and magnitude of temperature clearly corresponds with the underlying LCZ surface cover.

In the following examples, the potential for a consistent climate-based landscape classification scheme is illustrated for the ubiquitous urban effect on temperature (i.e., the UHI). But, of course, there are many other applications, such as air quality modeling, the creation of urban climatic maps to aid climate sensitive urban design (Ren et al. 2017), and improving the representation of cities in global climate models

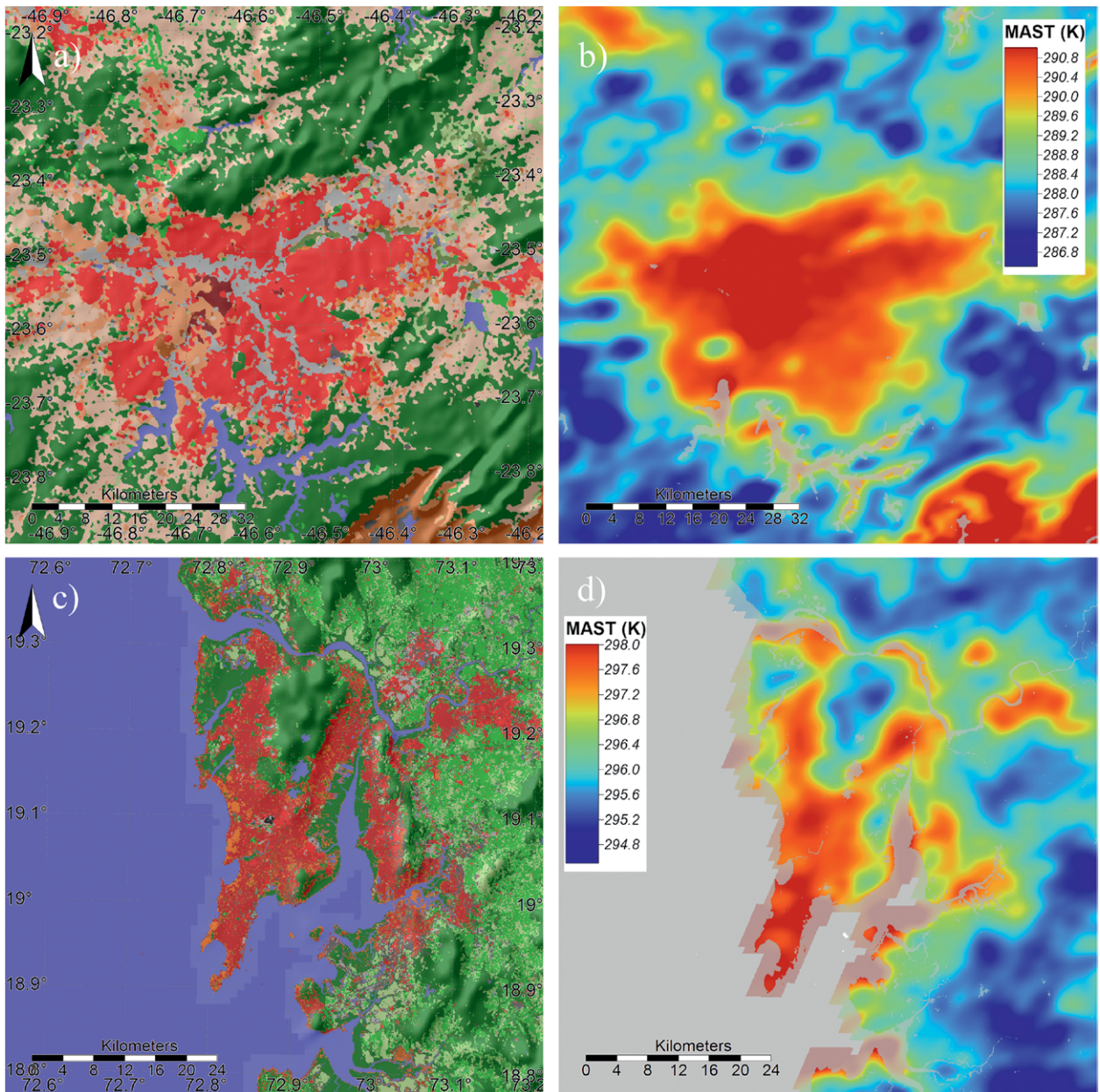


FIG. 5. WUDAPT L0 maps and MAST at 2230 LT in kelvin for (a),(b) São Paulo, Brazil, and (c),(d) Mumbai, India. The underlying topography shown in (a) and (c) is based on the Shuttle Radar Topography Mission. The LCZ legend is as in Fig. 2a.

(Feddema et al. 2015). The UHI, which includes the urban effect on surface, subsurface, and air temperatures, is one of the often-studied aspects of the urban climate. The surface UHI (UHI_{surf}) as observed from the vantage of a satellite (e.g., Fig. 5) and the near-surface (canopy level) UHI (UHI_{UCL}) are often used as measures of urban impact on building energy use and heat stress (Oke et al. 2017).

The cause of UHI_{surf} is primarily linked to the properties of construction materials (their radiative and thermal properties) and their dryness state—as

consequence, urban surfaces (when viewed from above) generally tend to be warmer by day and night (Oke et al. 2017). Therefore, the magnitude of the UHI_{surf} depends on both the character of the urban surface and the nature of the surrounding nonurban landscape (vegetative cover, moisture status, season, etc.). The UHI_{surf} can be simulated by solving the surface energy balance, which accounts for the exchanges of radiation and sensible and latent heat fluxes between the surface and the overlying atmosphere. The SUEWS model can derive these energy balance terms

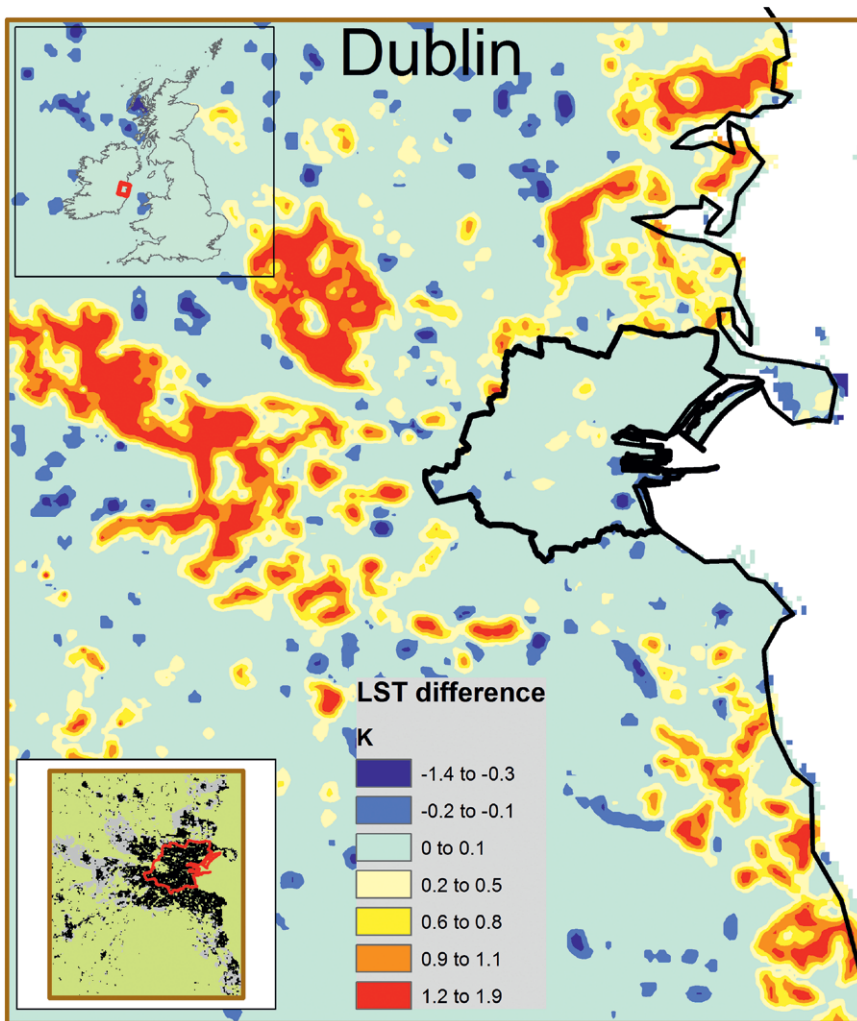


FIG. 6. The impact of urban growth on mean surface temperature for Dublin, Ireland (top inset map). The SUEWS model was run using parameter data derived from WUDAPT L0 data. Simulations were carried out for current (2006) and projected (2026) urban cover (bottom inset map) using typical Jun weather and the map shows the difference in their surface temperatures (K).

using commonly measured meteorological variables and information about land cover. For a given area it requires the fractional areas occupied by paving, buildings, coniferous trees/shrubs, deciduous trees/shrubs, irrigated grass, nonirrigated grass, and water. SUEWS has been evaluated across a range of urban landscapes and is ideally suited to simulate surface-air exchanges during weather dominated by clear and calm conditions that are conducive to UHI formation (Järvi et al. 2011).

Alexander et al. (2016) used SUEWS to examine the climate impacts of different urban development paths, using the example of Dublin, Ireland. Figure 6 shows the results of a simulation experiment, comparing the average surface temperature for June for Dublin in 2026, based on projections of population

the geometry and underlying material composition of the UCL that regulates the nighttime loss on longwave radiation (Oke et al. 2017) and the thermal character of the built fabric, which stores daytime heat and anthropogenic additions of heat. Atmospheric models that simulate the UHI_{UCL} require detailed information on the character of the urban canopy. The most sophisticated models will nest the microscale details of the urban canopy layer within larger-scale meso-scale processes that regulate the background climate.

Figure 7 shows the results of a study on the Madrid, Spain, UHI_{UCL} using WRF with the BEP-BEM scheme. The modeling setup consisted of five nested domains, with Madrid located in the inner domain of 7,200 km² (shown in the inset at the top of the figure) composed of 240 × 270 cells at a resolution of 333 m.

growth and urban growth made in 2006. Land cover in 2006 and 2026 was converted to LCZ types, which were then used to derive parameter values for SUEWS and simulations based on current climate. The results show a more extensive UHI_{surf} that reflects the replacement of natural surface cover by urbanization. On the other hand, an alternative projection based on increased building density rather than expanding the urban footprint does not change the UHI_{surf} appreciably. This experiment shows the potential value of WUDAPT data in an applied planning context.

The urban canopy layer UHI (UHI_{UCL}) describes the impact of cities on the near-surface (~2 m) air temperature; typically, the near-surface air in cities is warmer than that in the surrounding natural area and is strongest at night under clear skies and calm conditions in densely built parts of the city. Although it is linked to the UHI_{surf} , it has its own distinct genesis processes linked mostly to

The W2W tool generated the UCPs corresponding to the L0 maps that were used for the urban cells in the inner domain. Figure 7 shows the simulated surface air temperature under ideal weather conditions for UHI_{UCL} formation, which shows the correspondence between the urban footprint and the magnitude of the heat island. The model output using WUDAPT data was compared with output derived using data in the European Environment Agency's Urban Atlas (www.eea.europa.eu/data-and-maps/explore-interactive-maps/urban-atlas-for-europe), which has limited information on land cover within municipal boundaries. Observations made at a weather station network in the city provided an independent assessment of model performance. The results showed that performance of the model using L0 corresponding UCPs improved model performance by ~10% based on rmse and mean bias indicators. Given the relative ease with which LCZ maps can be generated, the results show the potential to greatly improve urban modeling capacity, particularly where no other land-cover data are available (Brousse et al. 2016).

Heat waves are a leading cause of weather-related fatalities globally, and there is evidence that the UHI can act synergistically with expected global climate change to enhance the risk to public health in cities (Li and Bou-Zeid 2013). In Fig. 8 the results of a study of heat stress in New Delhi, India, are presented. In this study a baseline event was simulated based on a heat wave event (22–27 May 2015) that advected very hot and dry air into the city; during this period the maximum and minimum temperatures in New Delhi reached 46° and 32°C, respectively. To examine the impact of urban growth on the intensity

and extent of the associated heat stress, L0 data were generated for the modeling domain at two time periods (1977 and 2015) and the W2W tool was used to generate appropriate UCP values for WRF (Niyogi et al. 2017). Simulations were performed using the synoptic forcing conditions that prevailed during the 2015 event and the National Oceanic and Atmospheric Administration (NOAA) heat index (HI) was calculated. HI represents the heat stress associated with high temperature and relative humidity as an “apparent” temperature; in Fig. 8 the difference between the HI values for 2015 and 1977 is presented. This difference map shows that urban development has increased both the spatial extent and the magnitude of the heat stress. This example illustrates the value of improved urban land-cover descriptions for extreme weather modeling predictions.

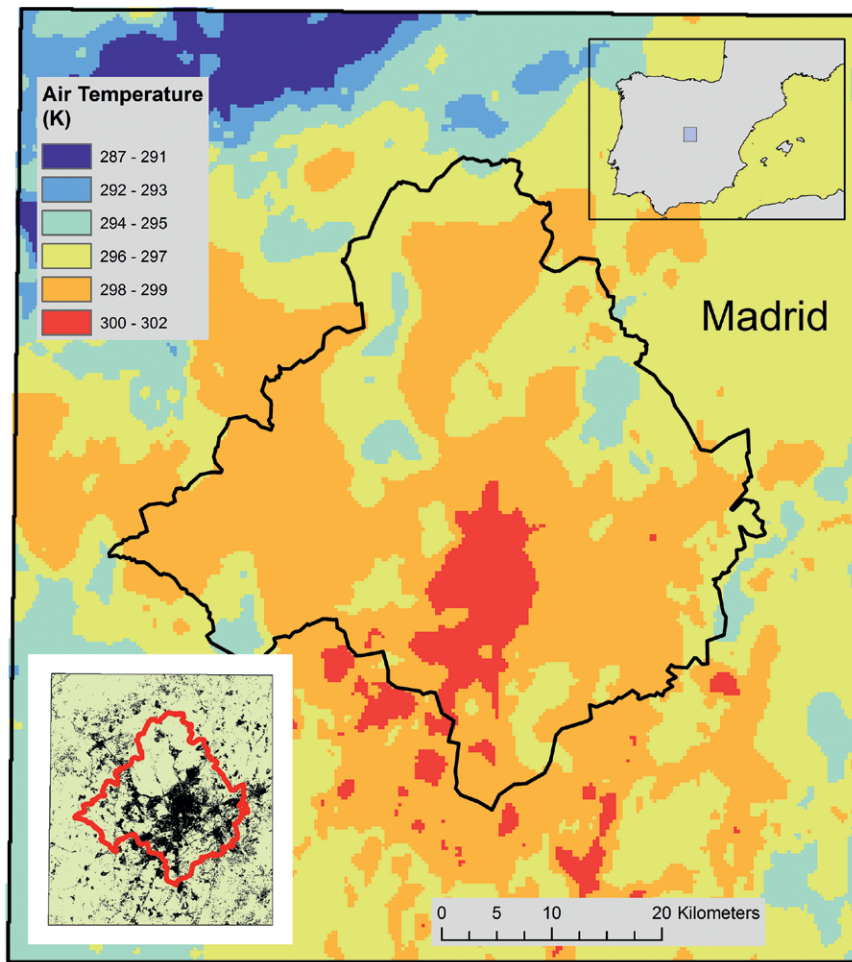


FIG. 7. The simulated near-surface air temperature (K) over Madrid (for the fifth nested domain shown in the top inset map) using the WRF Model. The urban canopy parameter values for WRF are from WUDAPT L0 data; the urbanized landscape is shown outlined in the bottom inset map. The map shows surface air temperature at 0300 LT 13 Jul 2015 during a heat wave event.

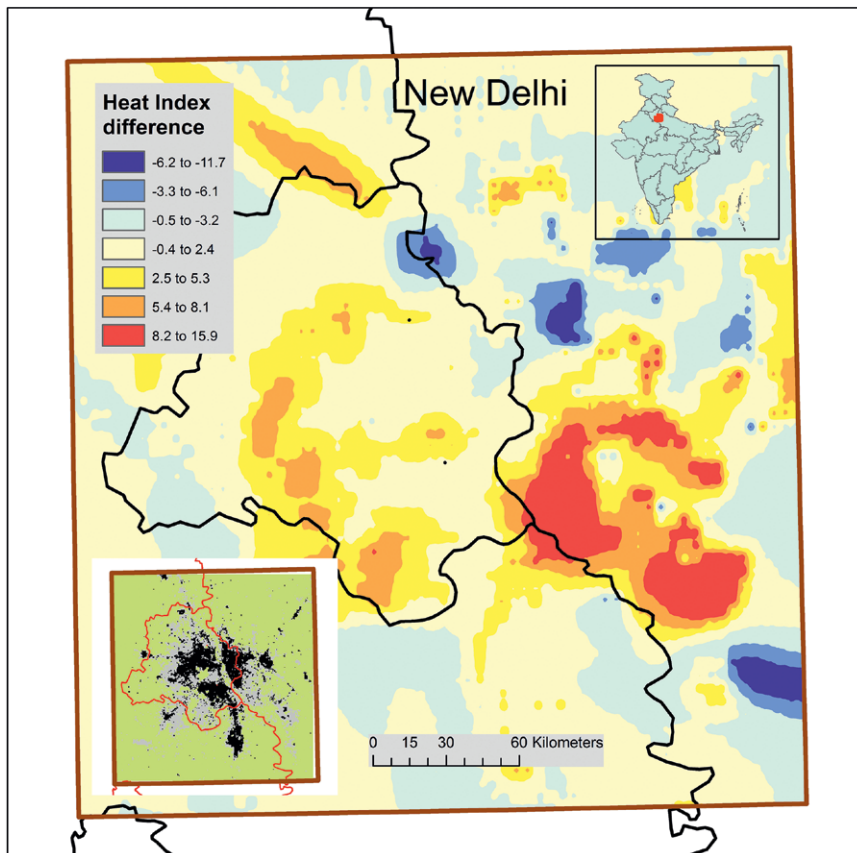


FIG. 8. The simulated impact of urban development on heat stress over New Delhi, India (top inset map). The WRF Model was run using WUDAPT L0 data for 1977 and 2015, based on weather conditions for 25 May 2015. The growth of the city over this period is shown in the bottom inset map. The WRF simulation was used to calculate the NOAA HI, expressed as apparent temperature in Fahrenheit; the figure depicts the difference: $HI_{2015} - HI_{1977}$.

CURRENT STATUS AND NEXT STEPS.

As it stands, researchers can use open-source tools and Landsat data to generate L0 data quickly, which overcomes a major obstacle to model application where there are no data currently. In addition to the projects presented above, which focused on the urban heat island, there is evidence that the dynamics and chemistry simulated in urban models are sensitive to the description of the underlying city surface. Figure 9 shows preliminary results from a study of air quality in Guangzhou, China, using the single-layer urban canopy model coupled to Noah in the WRF Model coupled with chemistry (WRF-Chem; Grell et al. 2005; Kusaka et al. 2001). The figure depicts the time–height cross section of simulated $PM_{2.5}$ (particles with size $\leq 2.5 \mu m$) distribution in the upper boundary layer for a fair weather period (15–17 October 2014) that corresponded with a pollution episode. The cross sections show two simulations based on a generic “urban” category (Fig. 9a) using

UCP values from Zhang et al. (2010a,b) and based on WUDAPT L0 data (Fig. 9b). The observable differences are the result of the simulated wind fields that reflect advanced urban physics parameterizations in WRF that can take advantage of the quality of urban data provided. Also, preliminary work on modeling air quality over São Paulo (Dirce et al. 2018, manuscript submitted to *Urban Climate*) confirms that significant spatial and temporal variability in the complex 3D flows and mixed layer height variations across the city are evident when more precise urban data (i.e., L0 data) are provided.

While WUDAPT continues to acquire L0 data for additional cities, the long-term strategy recognizes the need for a multidimensional approach to data gathering and processing with an emphasis on gathering additional socioeconomic

and surface variables. There are a number of activities underway to improve WUDAPT and its products and extend modeling application capabilities.

L0 data quality and UCP precision. Much of the effort in designing the protocol for L0 data has focused on ensuring the quality of the data. For example, experiments have demonstrated that using a contextual classifier that takes into account information in neighboring pixels during the LCZ mapping process can significantly improve the quality of the map (Verdonck et al. 2017). However, the quality of the TAs remains the foundation of the protocol for generating the LCZ maps. At a minimum, L0 data should be reproducible by independent evaluators to achieve a high level of self-consistency, but experience has shown that there is considerable variation among the urban experts in their creation of TAs. As part of the Human Influence Experiment (HUMINEX) initiative, Bechtel et al. (2017b) investigated 94

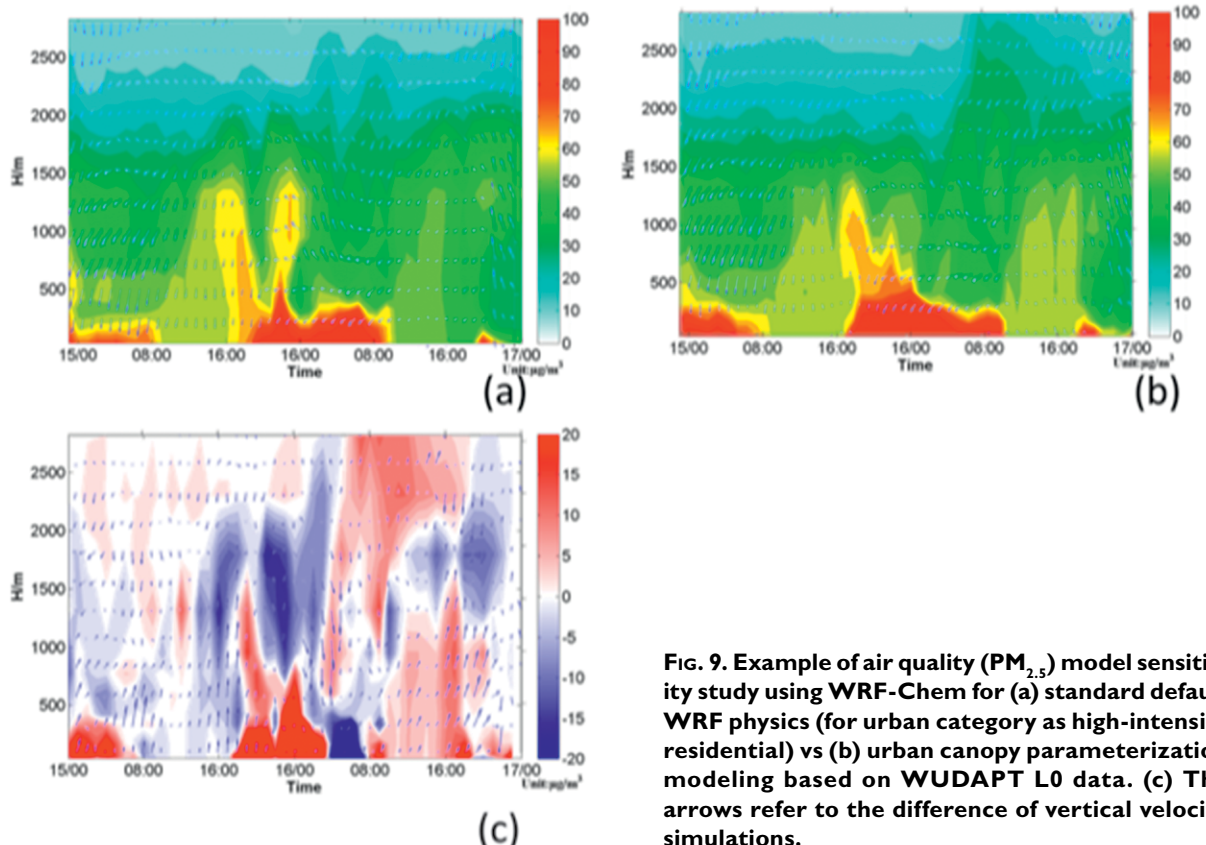


FIG. 9. Example of air quality ($PM_{2.5}$) model sensitivity study using WRF-Chem for (a) standard default WRF physics (for urban category as high-intensity residential) vs (b) urban canopy parameterization modeling based on WUDAPT L0 data. (c) The arrows refer to the difference of vertical velocity simulations.

crowdsourced training datasets for 10 different cities. The results indicate that while LCZ maps generated by TAs from one individual may be of poor quality, increasing the number of training data revisions and combining multiple training sets increases the quality of L0 data considerably.

In related work, cross evaluations are being undertaken with comparable urban land-cover information where it is available, such as the impermeable surface cover recorded in Europe's Urban Atlas and the built cover available in the Global Human Settlement Layer (Pesaresi et al. 2013). This work also has the potential to provide more precise UCP values for LCZ types, which are currently based on the information presented in Table 2. The objective of this endeavor is to generate guidance for assigning the most probable values of UCPs by LCZs to each grid in the modeling domain.

Actions to acquire higher-level data. Developing richer urban databases, in terms of both spatial detail and adding other relevant variables (such as building and vegetation characteristics), is a goal for the next phase of WUDAPT (Ching et al. 2017a). The information on buildings will be gathered using an approach similar to that for gathering L0 data, that

is, to develop and employ an international building typology with associated physical and functional properties. Data acquisition will rely on crowdsourcing techniques such as smartphone and web-based tools and will utilize the WUDAPT community (See et al. 2015). The paradigm for this initiative is based on the MAppUCE project, which employs France's building database to extract detailed UCPs related to building dimensions, construction materials, and occupation patterns (Masson et al. 2015, 2017). Members of the Passive Low Energy Architecture (PLEA) community are helping to create the WUDAPT building typology (Ching et al. 2017b). The existing L0 data (i.e., LCZ maps) will be used to provide a context for the data gathered and manage sampling across the urban landscape. The quality evaluation will require other independently derived data such as those available in some national censuses. Where possible, advanced satellite data and processing algorithms can provide high-definition data on building form (Wang and Dai 2015); the feasibility of this has already been demonstrated by Xu et al. (2017a,b). These sources could also provide UCPs, such as building volume density, ground coverage ratio, frontal area density, open spaces, and greenery coverage ratio.

Portal tools. WUDAPT tools are being developed to make maximum use of these data as they emerge. Priority capabilities under consideration include tools that 1) link the database to the wide variety of urban climate models in stand-alone configurations (e.g., Grimmond et al. 2010) or as components to larger-scale models such as WRF, 2) allow weather data gathered at World Meteorological Organization (WMO) standard stations outside cities to be transferred to urban locations (e.g., Erell and Williamson 2006), 3) combine with other available modeling software (e.g., Vanegas et al. 2012a,b) and land-cover data to create future urban growth scenarios, and 4) evaluate urban risks associated with current and future climate hazards, (e.g., Hanna et al. 2015).

Further enhancements may be possible through links enabled through WUDAPT (www.wudapt.org), given the rich datasets afforded by variety and types of remotely sensed datasets beyond the traditional and basic Landsat land-use–land-cover classifications. Inclusion of such information in WUDAPT would be as ancillary and auxiliary data for enhanced analyses (e.g., Comarazamy et al. 2013, 2015; Hulley et al. 2014; Imhoff et al. 2010; Luvall et al. 2015).

OUTLOOK. Urban issues are rapidly moving to the forefront of the challenges posed by climate changes across a hierarchy of scales. The WUDAPT project is developing a comprehensive global archive of urban data and associated tools that will be needed to address these challenges. The WMO is exploring the use of WUDAPT as a means toward addressing its new urban services mandates expressed in Resolution 68 (WMO 2015) and in development of the Guide for Integrated Urban Hydrometeorological, Climate and Environmental Services (Baklanov et al. 2018). In China, WUDAPT data have already been used for urban impact analyses studies of dynamic growth in the Pearl River delta (Ren et al. 2017) and in examining the impact of urbanization as part of China’s “One Belt, One Road” plan. WUDAPT is participating with the Group on Earth Observations (GEO) WUDAPT in the Global Human Settlement Layer Project (Pesaresi et al. 2013) and the Human Planet Initiative, focusing on activities associated with global urban climate and mitigation planning actions.

WUDAPT is a successful grassroots effort, and continued community involvement is key to assuring success. Please consider engaging in and/or following the progress online (www.wudapt.org).

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A Scientific Peak: How Boulder Became a World Center for Space and Atmospheric Science

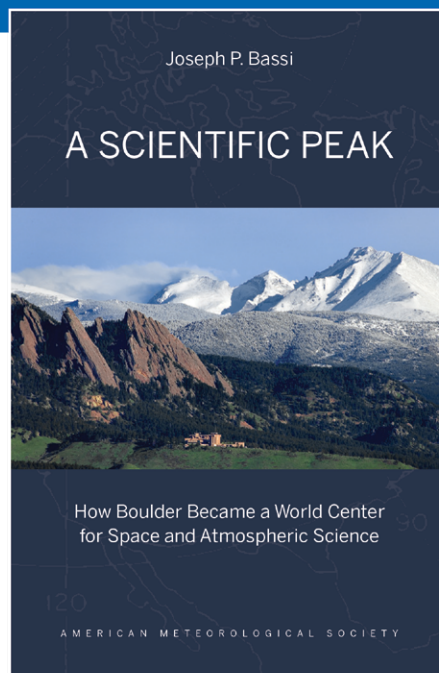
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