The author's response to report on The 'urban meteorology island': a multi-model ensemble analysis

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The author's response to Anonymous Referee #3

We would like to thank to Anonymous Referee #3 for all comments, suggestions and corrections in his review of our manuscript. We addressed all and below our point-by-point responses follow:

Referee's Comment #1: The fact that the results are achieved using a horizontal resolution of 9 km within the models needs to be mentioned more clearly in the abstract and conclusions.

Author's response: We agree with the reviewer, information about the horizontal resolution will be added to the abstract and conclusion. This is although a relatively coarse resolution, but the cities examined cover usually one or more model grid-boxes completely, moreover in the RegCM model, fractional urban land-use is considered which enables to include the effect of even minor urban areas.

Referee's Comment #2: The authors should also motivate why they have chosen the various combinations of physical schemes and possibly list some of the main physical characteristics of the respective schemes to improve readability.

Author's response: In general, the choosing of two models and various combinations of physical schemes is motivated by providing a very robust estimation of investigated changes. Considering the fact that we are interested mainly in local urban induced changes, we tested nearly all available schemes of urban canopies, boundary layer (connected with schemes of surface layer) and convection (for WRF model), and schemes of boundary layer, convection and microphysics (for RegCM). Convection parameterization can have potential effects on vertical transport of heat and moisture from the urban boundary layer while the model treatment of microphysics marks the hydrological budget over cities that influences the precipitation, latent heat release etc. The selection of different combination of model schemes is further based on their availability and restrictions in their different combinations.

Referee's Comment #3: Regarding the presentation of the urban results in section 3.2, a selection of those models who performed best in the overall evaluation would make much more sense.

Author's response: The general evaluation of model results in section 3.1 describes overall biases of temperature and precipitation, extended by comparison with station data for other variables. However, it does not serve as a validation of any urban-induced change, thus it does not tell us much about the accuracy of such changes. Moreover, it is not easy to say, which model is the best, from an overall view. E.g. simulations with the BEP+BEM urban model make higher biases in winter max/min temperature, but on the other hand, the wind speed is closer to reality (Fig. 4 and 5). Our study intends to provide a physical ensemble of the different components of UMI therefore, we present urban-induced changes of all simulations and/or averaged ones.

Referee's Comment #4: What aspects of the study have been surprising for the authors?

Author's response: Over the fact that the horizontal resolution is still relatively coarse (9 km), urban-induced changes (not only UHI) are clearly visible and mostly statistically significant. Especially the urban alteration of cloud cover and sub-grid scale precipitation in summer (Fig. 11 and 12), which are also described in observation-based studies. Further, the great influence of microphysics scheme (namely Nogherotto scheme in RegCM) on overall temperature and precipitation.

Referee's Comment #5: Can the authors also give a recommendation for the combination of physical packages using either WRF or RegCM?

Author's response: For our experiments with WRF that follow this study, we used the WU1L82C5 combination, e.i. SLUCM urban model, BouLac PBL and Eta SFC scheme, together with Grell-3D convection, as a less computational demanding compromise to BEP+BEM urban model, which needs more model levels. In terms of the RegCM model, probably the RUS simulation (UW PBL scheme, Tiedtke convection and SUBBEX microphysics scheme) makes the lowest biases in our domain.

The author's response to Anonymous Referee #1

We would like to thank to Anonymous Referee #1 for all comments, suggestions and corrections in his review of our manuscript. They cover issues which were resolved in the initial review, therefore they have been already taken into account by the authors and incorporated into the text. Nevertheless, we provide our point-by-point responses here too:

Referee's Comment #1: Abstract: the abstract should mention that you only study European cities. In terms of morphology they are substantially different North American cities that this caveat makes sense to mention.

Author's response: Corrected.

Referee's Comment #2: Abstract: you indicate that there is a substantial sensitivity of the model to the selected PBL schemes and urban canopy scheme. It would be more attractive for the reader if you can add a recommendation which settings are preferred.

Author's response: The primary task of the paper was not to provide an optimal model setup to describe urban climate but rather to examine the sensitivity of the modelled urban climate and its contrast with urban vicinity to different models and model configurations. Nevertheless, we added a such 'recommendation' to the discussion and conclusion based on Fig. 3, 4 and 5.

Referee's Comment #3–6: Ln 12: choice => selection, Ln 13, 31, 59: hyphenation: boundary-layer scheme. Please check the manuscript throughout.

Author's response: Corrected everywhere.

Referee's Comment #7: Ln 76: model resolution of 9 km: defend why this is sufficient to represent cities sufficiently. The resolved scales will be 5 * 9 km = 45 km, which means only cities of that scale are appropriately resolved, but there are not that many cities in Europe of that scale. I the feeling the cities on Belgium and Hungary are larger in your Fig. 1 than in reality.

Author's response: As mentioned in the manuscript, the urban land-use is differently represented in WRF and RegCM: in RegCM it is defined as fractional landuse which allows even small urban areas to be resolved as fractions of the 9km×9km grid-box. In WRF, grid-box is considered 'urban', when the urban land-use has the highest fraction, i.e. it does not have to be strictly over 50 % of grid-box area. Moreover, urban land-cover can be increased by small towns and villages located in the grid-box near the city, which seemingly increase the city size. However, big cities such as Berlin, Prague, Warsaw or Budapest, analysed in this paper, are covering multiple model grid-boxes themselves and thus are represented sufficiently even at 9km×9km resolution.

Referee's Comment #8: Ln 78: 2015–2016: please defend why these years have been selected. 2015 is a rather warm year in Europe, so how representative is the selected period.

Author's response: Indeed, 2015 is a warmer year than average (not so for 2016) (https://climate.copernicus.eu/european-temperature). However, years 2015 and 2016 are comparable to other years in the last decade (e.g. 2014, 2018) for Europe. Moreover, the variability of the local climate characteristic for individual cities chosen in the study is certainly larger (e.g. Hamburg with about 8 °C annual average temperature versus Beograd 12 °C; source: climate-data.org) than the year to year variability of the average European climate. We thus conclude that the spread of the magnitude of the urban meteorological effects and the 'urban meteorology island' given by considering a wide range of cities is well above the spread given by choosing different years during recent decade.

Referee's Comment #9: Ln 85: please add some sentences that defend why you have selected these schemes. I expect you have not selected them randomly but that there was a certain strategy or you built upon earlier studies.

Author's response: The selection is based on the availability of the schemes and the restrictions for their different combinations as well as on the expected impact on the 'urban meteorology island'.

For WRF: BEP+BEM urban model works only with MYJ and Boulac PBL schemes combined with the Noah land-surface model. The MYJ PBL scheme works only with Eta surface-layer scheme (see e.g. WRF-ARW user's guide). Further, schemes usable in WRF-Chem model with indirect aerosol effect are tested (Purdue Lin microphysics scheme) and those that enable

cumulus radiation feedback (listed convective schemes). These simulations are planned to be used in a following study with the radiative feedback of the emissions from the selected cities. The RRTMG radiation scheme was chosen as an efficient radiative transfer model that is fast yet has sufficient complexity.

For RegCM: two PBL schemes are available (Holtslag and UW) so these were altered as the PBL model description greatly influences the meteorology in the urban canopy. Further, the two most widely used convective schemes were tested, the Grell and Tiedtke schemes with the consideration that convection is an important process that removes heat and moisture from urban areas influencing urban climate. Finally, three microphysical schemes are available in RegCM (for the used version 4.7) and these were combined with the PBL and cumulus schemes. The urban canopy model could not be altered as only the CLM4.5/CLMU was available for RegCM simulations.

Referee's Comment #10: Table 1: elaborate the table caption. The caption should be placed above the table. Idem for Table 2 and 4.

Author's response: Changed for all tables.

Referee's Comment #11: Table 1: please add a sentence that elaborates on the experiment abbreviations. E.g. the 'E1U1L82C5' is not naturally related to the <SLUCM, 40, BouLac, Eta, Grell-3D> experiment. All experiment abbreviations start with 'E1' so E1 can be removed. Idem for Table 2.

Author's response: E1 is removed from all abbreviations, added 'W' on the beginning in Table 1, denoting WRF model, similarly to RegCM model in Table 2. Changed in all tables and figures, including the text.

Referee's Comment #12: Ln 118: simulations with MYJ PBL and Eta SFL schemes give notable underestimations (up to 2 deg C): link to literature, this MYJ behaviour is well known.

Author's response: This feature is commented and referenced (by Zhong et al., 2017) in the first paragraph of the Discussion section.

Referee's Comment #13: I have reservations against figure 3. Not about the contents but about the plot type. Now the results are shown as time series or at least the lines connect the different experiments. However there are no links between the connected experiments. So the results should be presented differently, e.g. as bar graphs. Idem for Fig 6.

Author's response: X-axis does not mean time axis automatically, lines between points are removed from the plot, points are enlarged (Fig. 3 and 6).

Referee's Comment #14: Ln 124: Daily ECAD values from selected stations are used for more detailed model validation over european urban areas, which are the main focus of the paper: European should be capitalized.

Author's response: Remark accepted.

Referee's Comment #15: Figure 4: please label all figures a,b,c, etc. This is much more easy for referencing.

Author's response: We believe that subplots are sufficiently denoted by variable names and statistical quantities, but in any case, the publisher places "a)", "b)" etc. subplot indicators, automatically during the typesetting process (including placement in the figures caption).

Referee's Comment #16: Figure 4: panel T2max and SWDOWN should have an y axis that is better adjusted (less wide range) to the data.

Author's response: Comment accepted. Figure 4 modified.

Referee's Comment #17: Figure 4: in columns 2 and 3 it is very difficult to see what are the differences between the runs. Better to start the y axis at a much higher value. Idem for Fig 5.

Author's response: Comment accepted. Figures 4 and 5 are modified.

Referee's Comment #18: Figure 4 and 5: if these statistics are averaged over all cities in Table 3, it remains unclear how they are influenced by certain sites or not.

Author's response: We are unsure if we understand the reviewer's comment properly. We tried to avoid a city-by-city validation to present rather the all-city-averages, which gives an indication of how the models are able to capture the urban climate for cities from an entire region. It is clear that models are more successful for some cities and less for others, but our aim is not to present individual city statistics.

Referee's Comment #19: Fig 8 and 9: the header "surface heat island" is misleading since this is not plotted according to the caption. Furthermore I do not understand what is the functionality of these plots if you only show satellite data.

Author's response: The header of Fig. 8 and 9 is "Surface urban heat island" as we plotted the surface temperatures (or skin-temperatures). The goal of presenting these two figures was to show the urban impact on surface temperature seen by remote sensing methods (satellites in this case). SUHI is a major component of what we call 'urban meteorological island' in our paper and we therefore decided to also show some observational based evidence of this component. Satellite measurements have the great advantage of showing the spatial distribution of the measured quantity in contrast with e.g. station based data.

Referee's Comment #20: Ln 198: please be more precise here: about which temperature are writing here. It is daily mean T2m?

Author's response: Yes, they are daily means of T2m during summer and winter season. The information has been added into the text.

Referee's Comment #21: Ln 202: absolute humidity is the density of the vapour pressure so unit should be g/m3.

Author's response: Following the Meteorological Glossary of American Meteorological Society, we changed it to "specific humidity", which describes mixing ratio, independent to adiabatic expansion or compression, and can be measured in $\mathrm{kg\,kg^{-1}}$ (i.e. it is dimensionless).

Referee's Comment #22: Ln 333: it is unclear whether the statement of BEP+BEM is an advertisement for this scheme or not. Is this scheme the best, despite the biases you report about?

Author's response: It is not easy to say whether the BEP+BEM is better or not – it produces greater biases in temperature but smaller biases in wind speed in comparison to other schemes.

Referee's Comment #23: Ln 349: I have the feeling the authors are somehow too positive about the satellite data. As far as I understand them, they can only be applied for cloud free days, and this does not occur very often, so they may give a biased picture. Please comment.

Author's response: Yes, surface temperature is measured only during clear-sky days by satellites, but over this fact, it still gives useful insight in surface temperature distribution around large cities, thus demonstrating the main component of the 'urban meteorology island' using remote sensing based data, which enables observation over larger areas rather than point measurements. The information about clear-sky days is added to the text.

The 'urban meteorology island': a multi-model ensemble analysis

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Abstract.

Cities and urban areas are well-known for their impact on meteorological variables and thereby modification of the local climate. Our study aims to generalize the urban-induced changes of specific meteorological variables by introducing a single phenomenon – the urban meteorology island (UMI). A wide ensemble of 24 model simulations with the WRF and RegCM regional climate models on European domain with 9 km horizontal resolution was performed to investigate various urban-induced modifications as individual components of the UMI. The results show that such an approach is meaningful, because nearly in all meteorological variables considered, statistically significant changes occur in cities. Besides previously documented urban-induced changes of temperature, wind speed and boundary-layer height, the study is focused also on changes of cloud cover, precipitation and humidity. An increase of cloud cover in cities, together with a higher amount of sub-grid scale precipitation is detected in summer afternoons. Specific humidity is significantly lower in cities. Further, the study shows that different models and parameterizations can have a strong impact on discussed components of UMI. Multi-layer urban scheme with anthropogenic heat considered increases winter temperatures by more than 2 °C and reduces wind speed more strongly than other urban models. Also the selection of planetary boundary-layer scheme influences the urban wind speed reduction, as well as boundary-layer height with the greatest extent. Finally, urban changes in cloud cover and precipitation are mostly sensitive to the parameterization of convection.

1 Introduction

Climate is one of the most important factors that influences the conditions for life at a specific place. Considering the fact that half of current population lives in cities (United Nations, Department of Economic and Social Affairs, 2014) and the population of largest cities is still projected to increase (Baklanov et al., 2016), and that the total number of cities of different sizes is rising (Mirzaei, 2015), the general knowledge of typical urban climate features and the difference with respect to their rural counterparts is becoming more and more crucial.

The most well-known urban climate feature is the so-called urban heat island (UHI), firstly described several decades ago (Oke and Maxwell, 1975), which means, in simple terms, that the urban temperatures are higher compared to rural ones. In

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following years, a large number of UHI observations was performed (e.g. Oke, 1982; Godowitch et al., 1985; Wolters and Brandsma, 2012; Theeuwes et al., 2015) and several empirical relation for UHI intensity (difference of city centre and vicinity temperature) were introduced (Oke, 1982; Theeuwes et al., 2017). More recently, once the computational power enabled a finer grid resolution, many of modelling studies, focusing primarily on UHI were performed (e.g. Ryu et al., 2013; Huszar et al., 2014; Trusilova et al., 2016; Göndöcs et al., 2017; Karlický et al., 2018; Huang et al., 2019). Specialized models of urban canopies within standard numerical weather prediction and regional climate models were used to capture the specifics of urban climate features.

However, several observation and model-based studies show that other meteorological variables are significantly altered by urban canopies too. The impact of cities on boundary-layer structure was documented already many years ago (Godowitch et al., 1985; Oke, 1987; Angevine et al., 2003), similarly to the impact on wind flow (Oke, 1987; Klein et al., 2001; Droste et al., 2018). These modifications of atmospheric dynamics over cities have considerable consequences to mixing, dispersion of pollutants and air quality in urban areas, which is also confirmed by model studies investigating the urban-induced changes of dispersion conditions (Karlický et al., 2018) and of primary or secondary pollutant concentrations (Fallmann et al., 2016; Huszár et al., 2018; Huszar et al., 2018; Li et al., 2019; Huszar et al., 2020; Ďoubalová et al., 2020).

Recently, some studies have investigated the impact of urban canopies on convection, cloud cover and rainfall. E.g. Theeuwes et al. (2019) describe the observed cloud cover enhancement over Paris and London during summer as a consequence of increased convection caused by UHI. Manola et al. (2020) show increased summer precipitation intensities and also overall precipitation increase in the city of Amsterdam in comparison to rural surroundings, also as a result of enhanced convection over the urban area. Enhanced convection, turbulence and mixing can cause, under special weather conditions, even wind speed increase over cities (Droste et al., 2018). Finally, humidity is also impacted by urban canopy, e.g. Langendijk et al. (2019) show a significant decrease of relative humidity in Berlin (primarily in summer) and a less expressed decrease in specific humidity.

All these urban-induced meteorological changes, forming island-like features in the spatial distribution of the mentioned variables over urban areas, resulted in a formulation of new concepts describing the urban weather and climate. Apart from the UHI (Oke and Maxwell, 1975), other meteorology-related "urban islands" were defined recently: urban dry island (UDI; Moriwaki et al., 2013), urban cool island (UCI; Theeuwes et al., 2015), surface urban heat island (SUHI; Göndöcs et al., 2017) and urban wind island (UWI; Droste et al., 2018). Such a relatively wide range of well documented island-like perturbations of physical fields over urban areas in contrast to their rural counterparts basically means that the meteorological conditions differ due to changes of the whole atmospheric physics and dynamics in urban areas. Therefore, we introduce here a generalization of the above mentioned urban area islands: the urban meteorology island (UMI) phenomenon, which we consider as an urban area that has significantly different meteorological conditions from surrounding rural areas. The specific "one-variable" islands like UHI, UDI, UCI, SUHI and UWI can be regarded as components of UMI, denoted as UXI, in general.

Considering the above-mentioned modelling studies, models well simulate, at least qualitatively, most of the processes leading to the UHI and other elements of the UMI are also simulated with all expected features. However, large differences exist in the magnitude of individual elements of the UMI between different models and their configurations (e.g. Trusilova et al., 2016; Karlický et al., 2018; Huang et al., 2019; Huszar et al., 2020). It is clear that the resulting UMI will be highly

dependent on how the relevant UXI-forming processes are represented in models, which include parameterization of sub-grid
60 processes as boundary-layer turbulence, convection, air-surface heat and water exchange, micro-physics and urban canopy
physics. Also the choice of driving model itself can have a potentially strong impact.

Motivated by this, here, we present a novel study that 1) perceives the urban-induced meteorological changes as one UMI
concept where urban perturbations of specific variables (UXI) are considered as components of the whole UMI. Further, 2) a
large number of cities from different locations in a model domain is taken into account to enable a robust, city-independent
65 view on the impact of a specific model and parametrization on the resulting UMI. Next, 3) a wide range of the model simulation
ensemble provides a robust estimation of different components of UMI, including their magnitudes and temporal evolution.
Finally, 4) a multi-variable validation of the whole model ensemble is performed here that brings an useful view for other
model users and can be feedback for model developers.

The paper is composed as follows: after the Introduction, the models, the data and the design of the experiments is presented,
70 followed by the Results section that contains a detailed validation of different model set-ups and the comparison of UMI
components. Finally, the results are discussed and conclusions are drawn.

2 Models and data

2.1 Models used

To achieve results of the study, several model simulations were performed by involving two meteorological models, namely
75 the Weather Research and Forecasting (WRF; Skamarock et al., 2008) model in version 4.0.3 and Regional Climate Model
(RegCM; Giorgi et al., 2012) in version 4.7. In both cases, the ERA-interim data (Dee et al., 2011) were used as a driving
meteorology. The computational domain was also the same for both models, specifically a 190×166 domain with 9 km hor-
izontal resolution over Europe, centred over Prague, Czech Republic (Fig. 1, Fig. 2). The simulation time range is 2 years
(2015–2016), with December 2014 as spin-up. Static geographic data are taken from standard WRF and RegCM input, only
80 land-use fields are derived from CORINE Land Cover data, version CLC 2012. While WRF uses a dominant land-use (i.e.
one land-use type for a particular grid-box, Fig. 1), RegCM works with a fractional land-use (more land-use types included
proportionally in one grid-box, Fig. 2). Urban canopy parameters are the same as in Karlický et al. (2018).

The selection of model schemes is based on their availability, restrictions in their different combinations, expected impact on
the urban effects and to enable cumulus radiation feedback. In all WRF simulations, the radiative transfer is parameterized by
85 the Rapid Radiative Transfer Model for General Circulation Models (RRTMG; Iacono et al., 2008), microphysical processes
are resolved by the Purdue Lin scheme (Chen and Sun, 2002) and land-surface exchange by the Noah land-surface model (Chen
and Dudhia, 2001). The specific WRF simulations differ in parameterizations of boundary-layer processes, which are resolved
by the Mellor–Yamada–Janjic scheme (MYJ; Janjić, 1994) or the BouLac PBL scheme (Bougeault and Lacarrere, 1989) and
in surface layer processes (SFL) description, using the scheme as in Eta model (Janjić, 1994) or the Revised MM5 scheme
90 (Jiménez et al., 2012). Further, multi-layer Building Environment Parameterization (BEP; Martilli et al., 2002) linked to the
Building Energy Model (BEM; Salamanca et al., 2009), the Single-Layer Urban Canopy Model (SLUCM; Kusaka et al., 2001)
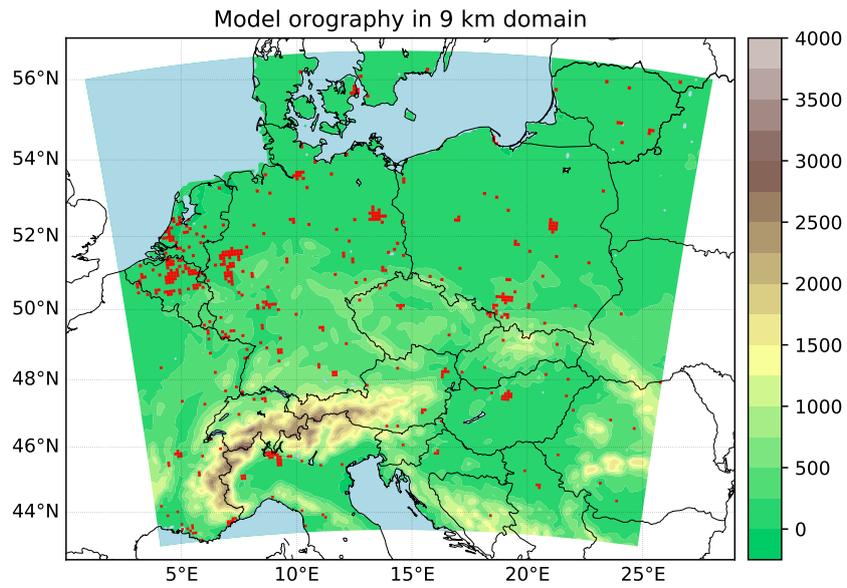
Figure 1. Position of model domain with model orography (m) and grid-boxes marked with red, in which the urban land-use type is dominant.
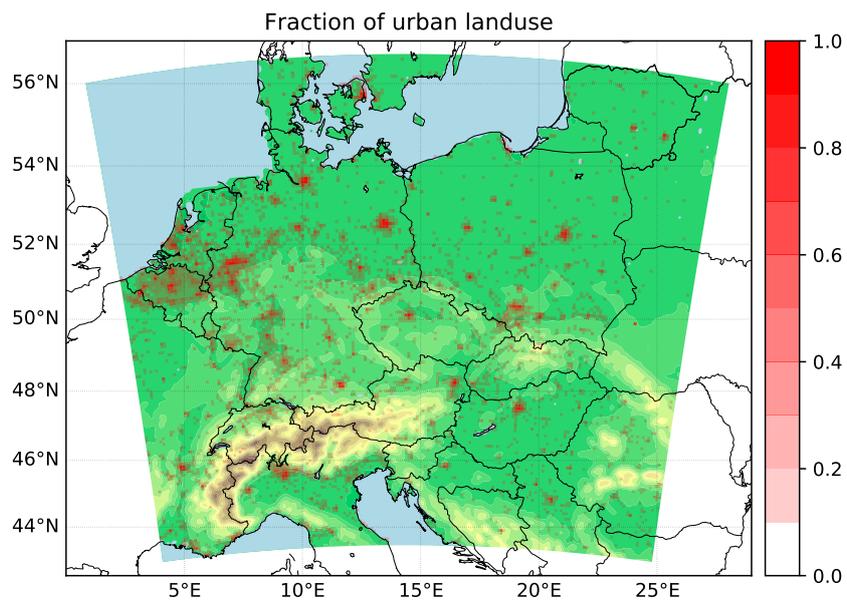

Figure 2. As Fig. 1 but with red marked urban land-use proportion.

Table 1. WRF model setups.

Experiment	Urban model	# levels	PBL scheme	SFL scheme	Convection
WU3L22	BEP+BEM	49	MYJ	Eta	Grell-Freitas
WU3L82	BEP+BEM	49	BouLac	Eta	Grell-Freitas
WU3L81	BEP+BEM	49	BouLac	MM5	Grell-Freitas
WU1L22	SLUCM	40	MYJ	Eta	Grell-Freitas
WU1L22C5	SLUCM	40	MYJ	Eta	Grell-3D
WU1L22C1	SLUCM	40	MYJ	Eta	Kain-Fritsch
WU1L82	SLUCM	40	BouLac	Eta	Grell-Freitas
WU1L82C5	SLUCM	40	BouLac	Eta	Grell-3D
WU1L82C1	SLUCM	40	BouLac	Eta	Kain-Fritsch
WU1L81	SLUCM	40	BouLac	MM5	Grell-Freitas
WU1L81C5	SLUCM	40	BouLac	MM5	Grell-3D
WU1L81C1	SLUCM	40	BouLac	MM5	Kain-Fritsch
WU0L22	bulk	40	MYJ	Eta	Grell-Freitas
WU0L82	bulk	40	BouLac	Eta	Grell-Freitas
WU0L81	bulk	40	BouLac	MM5	Grell-Freitas

and the bulk parameterization are tested for processes in urban environment. Convection is parameterized by the Grell-Freitas (Grell and Freitas, 2014), Grell-3D (Grell, 1993) and Kain-Fritsch scheme (Kain, 2004). In total, 15 specific combinations of chosen schemes were used and they are described in Table 1.

95   In terms of RegCM setup, all simulations are run on 40 model layers, the NCAR Community Climate Model Version 3 (CCM3; Kiehl et al., 1996) is used to parameterizing of the radiation transfer, land-surface transfer is resolved by the CLM4.5 (Lawrence et al., 2011; Oleson et al., 2013) scheme and urban-canopy-layer processes are solved by the Community Land Model Urban (CLMU; Oleson et al., 2008) scheme. Here, the model ensemble is created by two different parameterizations of boundary-layer processes – the University of Washington (UW; Grenier and Bretherton, 2001; Bretherton et al., 2004) and the

100 Holtslag (Holtslag et al., 1990) PBL scheme, by two different approaches of solving sub-grid convection – the Tiedtke scheme (Tiedtke, 1989) and the Grell scheme (Grell, 1993) and by three schemes of microphysical processes (MP) – explicit one-moment scheme by Nogherotto et al. (2016), older SUBBEX scheme (Pal et al., 2000) and the explicit 5-class single moment WSM5 model (Hong et al., 2004). The specific combinations of chosen schemes in terms of RegCM model are described in Table 2, the first model setup is considered as a baseline setup.

105 **2.2   Validation data**

To assess the model biases for temperature and precipitation over the entire domain, the E-OBS dataset (Haylock et al., 2008) of version 17.0 is chosen. Further, station-based data from European Climate Assessment and Dataset (ECAD; Klein Tank

Table 2. RegCM model setups.

Experiment	PBL scheme	Convection	MP scheme
RU	UW	Tiedtke	Nogherotto
RUHo	Holtslag	Tiedtke	Nogherotto
RUG	UW	Grell	Nogherotto
RUS	UW	Tiedtke	SUBBEX
RUW	UW	Tiedtke	WSM5
RUHoW	Holtslag	Tiedtke	WSM5
RUGW	UW	Grell	WSM5
RUHoS	Holtslag	Tiedtke	SUBBEX
RUGS	UW	Grell	SUBBEX

Table 3. Used station from ECAD for specific variables.

T2max/T2min	Cloud cover	Humidity	SW radiation	Wind speed
Dresden-Klotzche	Dresden-Klotzche	Dresden-Klotzche	Dresden-Klotzche	Dresden-Klotzche
Wien-Hohe Warte	Wien-Hohe Warte	Veliki Dolenci	Grossenzersdorf	Veliki Dolenci
Budapest	Hurbanovo	Novi Sad	Timisoara	Hurbanovo
Beograd	Beograd	Beograd	Craiova	Nove Mesto
Zagreb-Gric	Zagreb-Gric	Zagreb-Gric	Ostrava-Poruba	Celje
Warszawa-Okecie	Kosice	Gorlitz	Belsk	Kosice
Berlin-Dahlem	Berlin-Dahlem	Berlin-Dahlem	Potsdam	Berlin-Dahlem
Hamburg-Fuhlsbuettel	Hamburg-Fuhlsbuettel	Hamburg-Fuhlsbuettel	Hamburg-Fuhlsbuettel	Hamburg-Fuhlsbuettel
Muenchen	Muenchen-Stadt	Muenchen	Nurnberg	Muenchen
Falsterbo	Falsterbo	Falsterbo	Schleswig	Falsterbo

et al., 2002) are used for more detailed validation of model results in terms of daily maximum and minimum temperatures, cloud cover, relative humidity, downward shortwave radiation and wind speed. For every variable, ten stations were chosen that are equally-distributed over the domain. The specific stations are listed in Table 3.

Model surface (skin) temperature is validated by data from the Moderate Resolution Imaging Spectroradiometer (MODIS), operated by Terra and Aqua satellites (Wan et al., 2015a, b). For further use, monthly values of surface temperature in $0.05°$ horizontal resolution are selected. These are computed only from observations with clear-sky conditions, when satellite sensors are able to scan the earth surface. In our area of interest, four observations per day are performed, approximately at 10 and 21 UTC by the Terra satellite and at 2 and 12 UTC by the Aqua satellite.

3   Results

3.1   Model validation of model simulation ensemble

First, the general comparison of model temperatures and precipitation against E-OBS data is performed. Seasonal temperature and rainfall sums are averaged over the whole domain (excluding twenty rows and columns at the domain edges) and compared with corresponding E-OBS data. Results are shown in Fig. 3. Most of the WRF simulations predict average temperatures correctly, only simulations with MYJ PBL and Eta SFL schemes give notable underestimations (up to 2 °C). RegCM simulations with the Nogherotto MP scheme produce substantial biases, mainly in spring and summer season. Precipitation biases are related mainly to the chosen convection scheme: for WRF, the average summer overestimation is partly reduced by using the Grell-3D scheme. In RegCM, the Nogherotto MP scheme influences precipitation (besides temperature), resulting in a high bias during the whole year. Further, it seems that experiments with the Grell convection scheme are marked with higher summer overestimation than those with the Tiedtke convection.
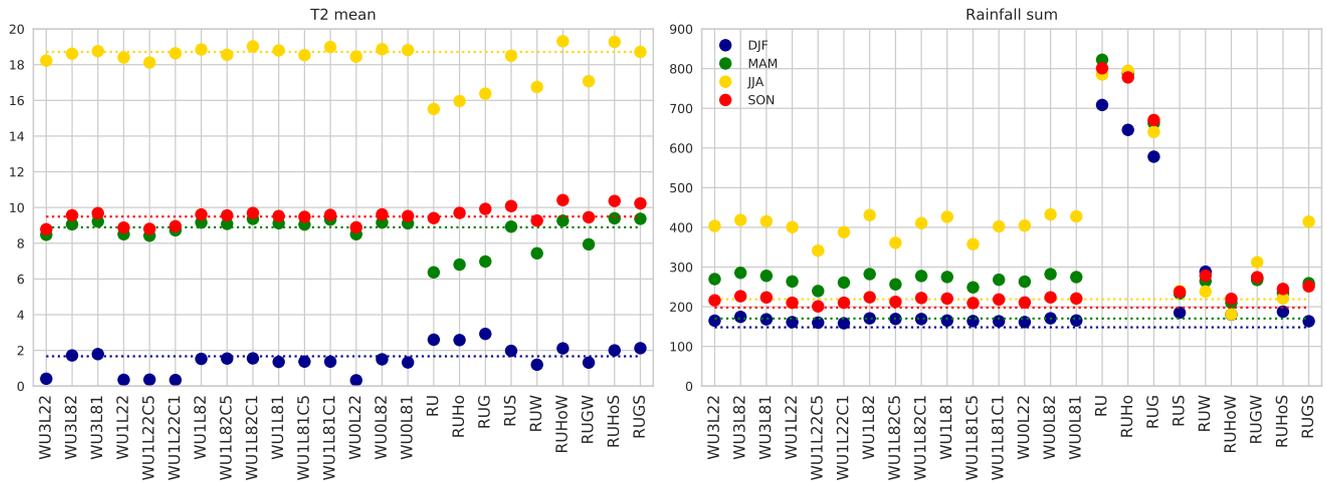
Figure 3. Domain averaged seasonal temperatures (°C) and precipitation sums (mm) for individual simulations and E-OBS data (dotted lines).

Daily ECAD values from selected stations are used for more detailed model validation over European urban areas, which are the main focus of the paper. Bias, correlation coefficient and regression coefficient (slope) are computed for all model simulations between ECAD and model time-series including values for every day of specific season and for all stations. The results are presented in Fig. 4 (winter season) and Fig. 5 (summer season).

Because measuring stations are located mainly in cities (Table 3), the type of the urban canopy parameterization impacts the results too. E.g. the BEP+BEM urban canopy model gives temperature extremes clearly higher than SLUCM or bulk and with bias up to 2 °C. Again, the combination of MYJ PBL and Eta SFL schemes gives lower temperatures than other combinations, which mostly means a negative temperature bias. The remaining WRF simulations predict temperatures reasonably well. For
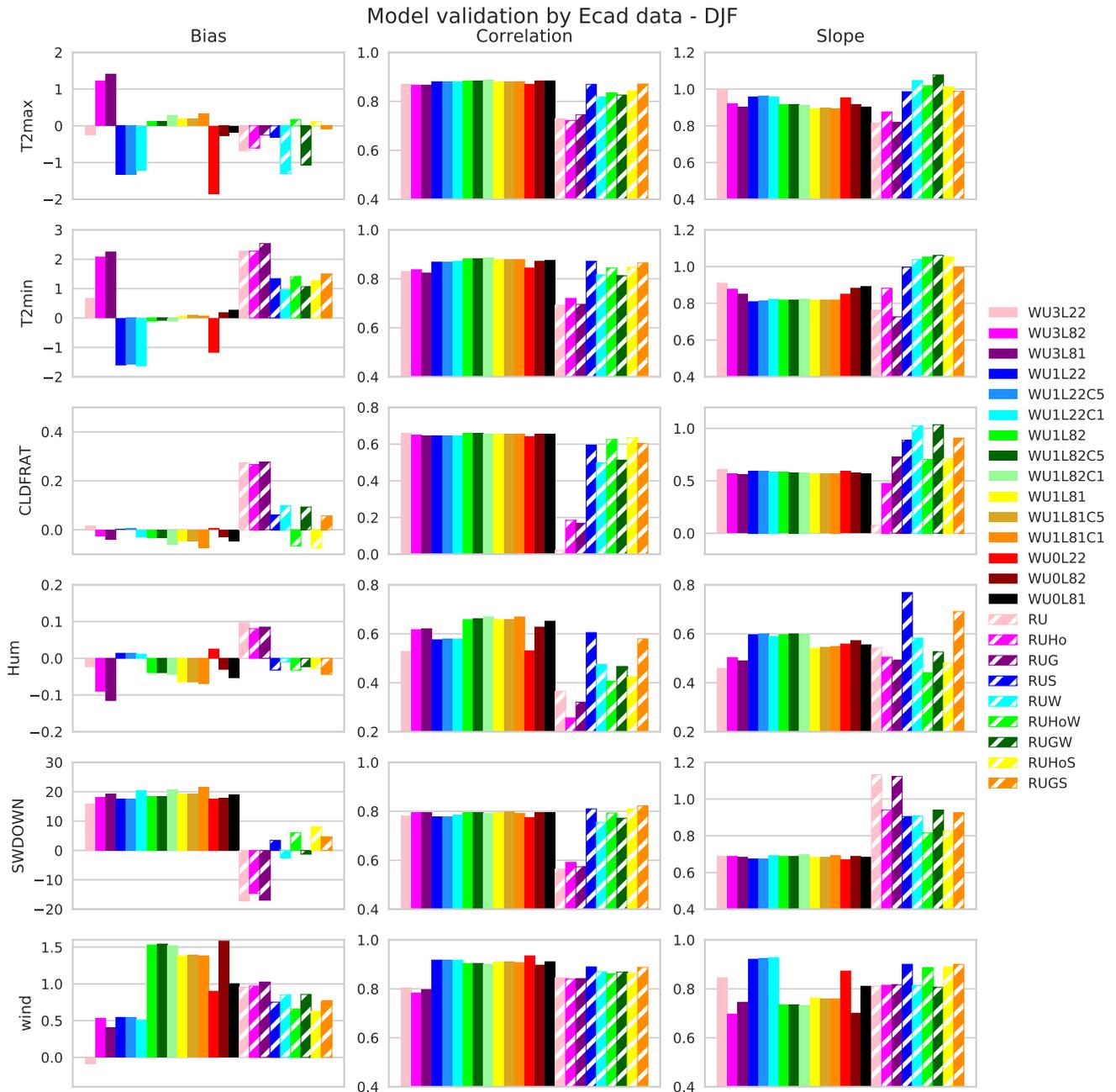
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Figure 4. Statistical evaluation of individual simulations using ECAD station data for selected variables in winter season. Units are $^\circ$C (temperature), fractions of one (cloud cover and relative humidity), $W\,m^{-2}$ (downward shortwave radiation) and $m\,s^{-1}$ (wind speed).
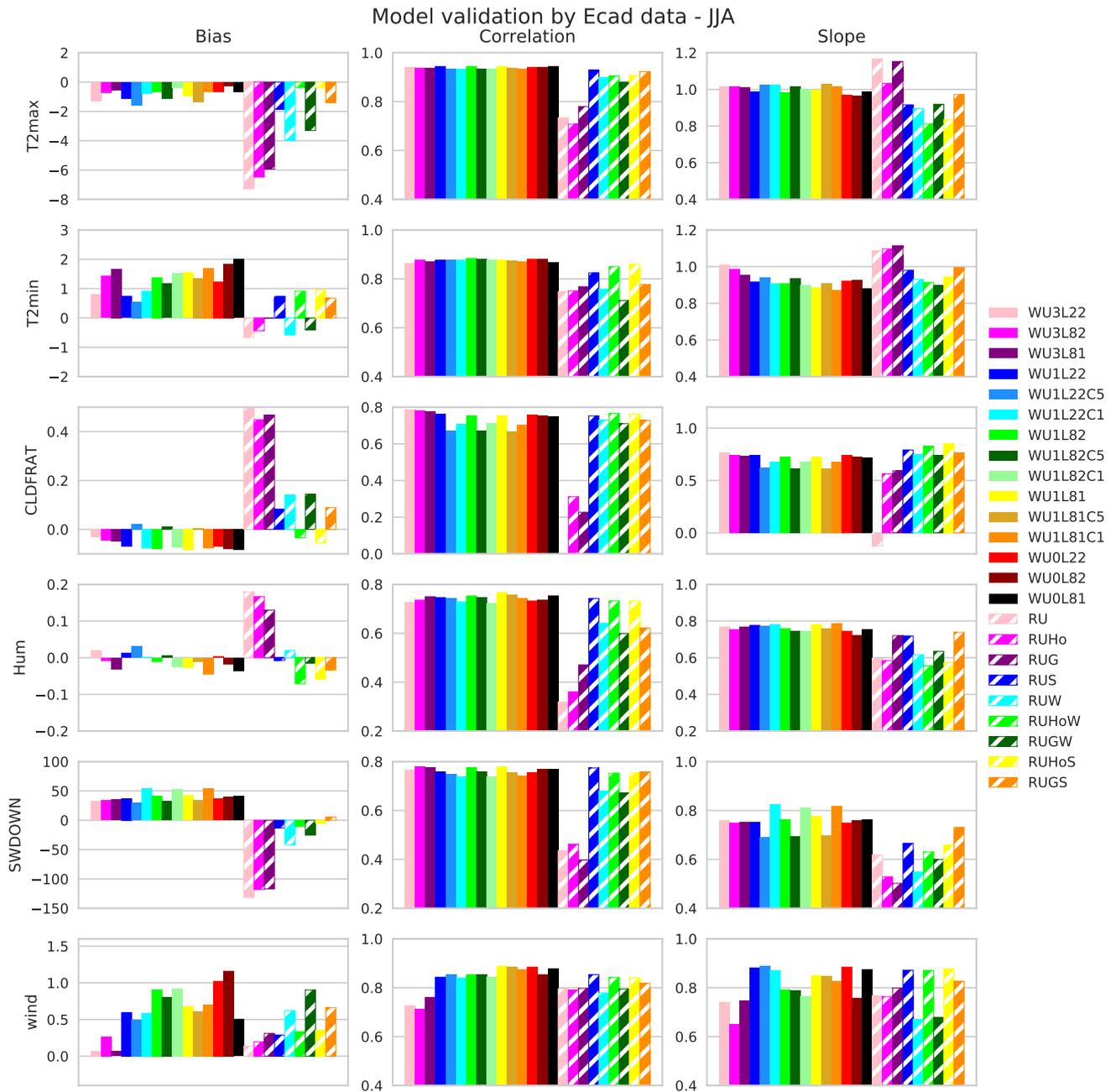
Figure 5. Same as Fig. 4 but for summer season.

RegCM, simulations with the Nogherotto MP scheme exhibit highest biases in minimum temperatures, as well as in case of cloud fraction. The biases of humidity and radiation are less than 0.1 and 20 W m^{-2}, respectively. Wind speed is overestimated by models, but the biases vary significantly between specific models and setups; the BEP+BEM urban model and simulations with the MYJ PBL and Eta SFL schemes, as well as all RegCM simulations, make a less bias. Correlations of temperatures, radiation and wind speed are, in general, around 0.8 or higher. On the other hand, for cloud fraction and humidity, they are only about 0.6. However, RegCM simulations with Nogherotto MP have the correlation substantially lower (except wind speed).

In summer, again, RegCM simulations with Nogherotto MP are marked with highest biases and the lowest correlations (with exception of minimum temperature and wind speed). Further, WRF simulations predict temperatures with biases less than 2 °C, while biases of some other RegCM simulations are up to 4 °C in case of maximum temperature. Biases of cloud fraction and humidity are mostly less than 0.1 and the sign of cloud fraction bias is clearly linked to the sign of radiation bias. WRF simulations are characterized by positive radiation bias up to 50 W m^{-2}. In the case of RegCM, a negative radiation bias prevails. As in winter, simpler urban models produce higher wind speeds in cities that results in higher positive biases. Biases of RegCM model for wind speed depend mainly on choice of PBL scheme. Correlations between model and observed values are of similar magnitude as in winter.

Being an important parameter of the urban canopy thermodynamic state, we also compare the modelled surface (skin) temperatures with satellite-based data. With consideration of the fact that satellite-based data are available only in specific times of day, roughly at 2 UTC, 12 UTC (Aqua satellite), 21 UTC and 10 UTC (Terra satellite), and the comparison of both daytime and nighttime satellite data with corresponding model values shows very similar features, averaging over both satellites is performed to get simple daytime and nighttime satellite data of surface temperature. Seasonal and domain averages of daytime and nighttime data together with the corresponding model simulation results are displayed in Fig. 6. Considering the fact that only clear-sky days are included in satellite-based data, Fig. 6 offers a rough comparison of modelled and observed values. In general, summer surface temperatures are predicted more accurately than winter ones, where model values are significantly higher (by 2–4 °C). Analogously to temperature biases against E-OBS (Fig. 3), the highest deviations are detected in RegCM simulations with the Nogherotto MP scheme, mainly in summer daytime (exceeding 5 °C). Again, WRF simulations with the MYJ PBL and Eta SFL schemes give surface temperatures up to 2 °C lower than remaining WRF simulations. Modelled and observed surface temperatures are consistent in feature that daytime spring values are higher than autumn ones, while nighttime spring surface temperatures are lower than autumn ones.

3.2 Components of the Urban Meteorology Island

As the most important component of the UMI, we show here the air temperature alteration over urban areas. Because of many previous studies describing the classical UHI, we show only a multi-model average of temperatures for chosen big cities over the domain for winter and summer season (Fig. 7). The UHI is clearly visible with a magnitude of about 1 °C in winter and 1.5 °C in summer, even using a relatively coarse model resolution (9 km).

Similarly, surface (skin) temperature is significantly altered by urban surfaces, which is obvious from satellite-based measurements of surface temperature, during clear-sky conditions. The same procedure as in the previous section was used to

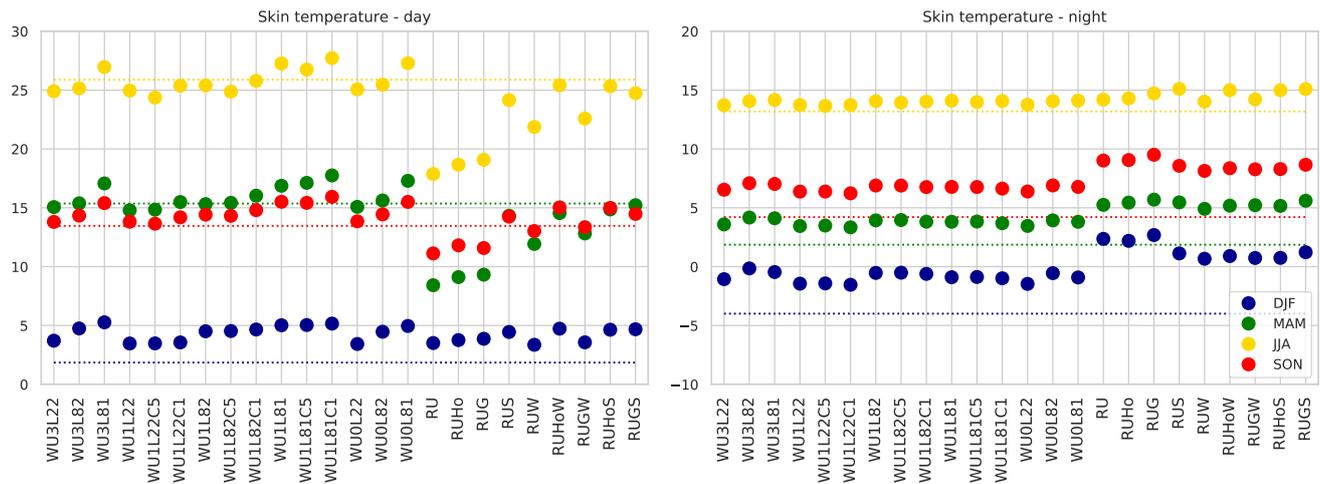**Figure 6.** Averaged seasonal whole-domain surface temperatures (°C) in daytime and nighttime given by specific simulations and satellites (dotted line).
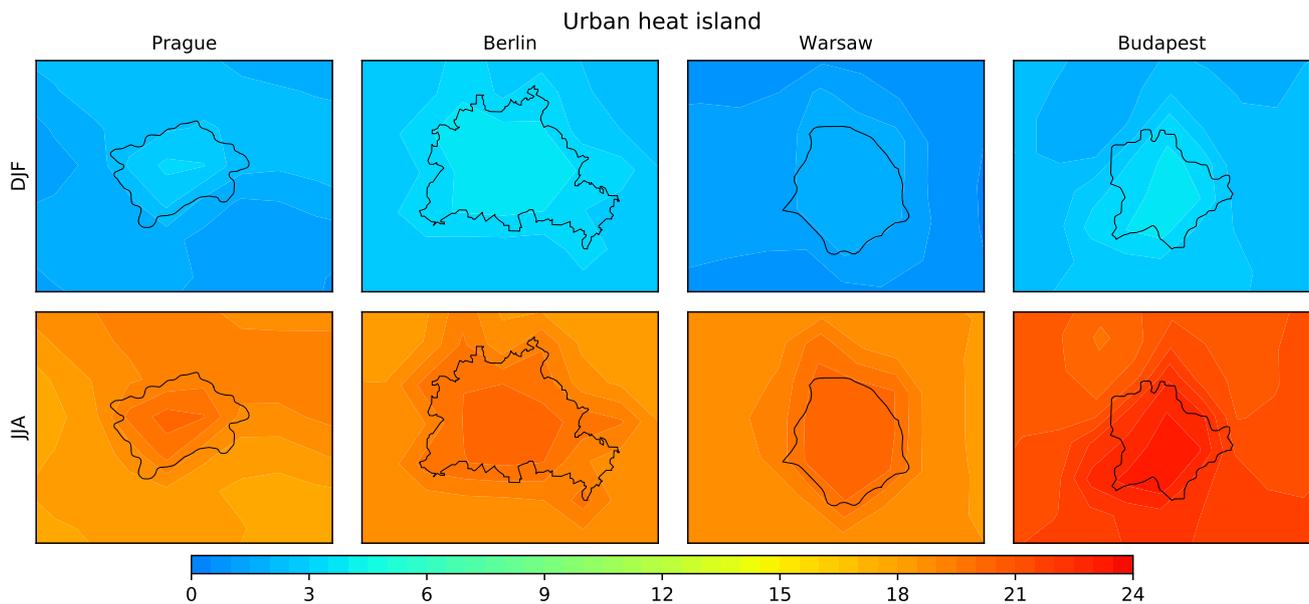

Figure 7. Multi-model averaged winter and summer temperatures (°C) around Prague, Berlin, Warsaw and Budapest

determine daytime (Fig. 8) and nighttime (Fig. 9) values. SUHI is most pronounced in summer daytime, when urban temperatures are approximately 4 °C higher. During nighttime, the SUHI intensity is slightly smaller, about 3 °C. In winter, SUHI is not so clearly pronounced as in summer and reaches 2 °C, except for Warsaw and Budapest where it is very small.
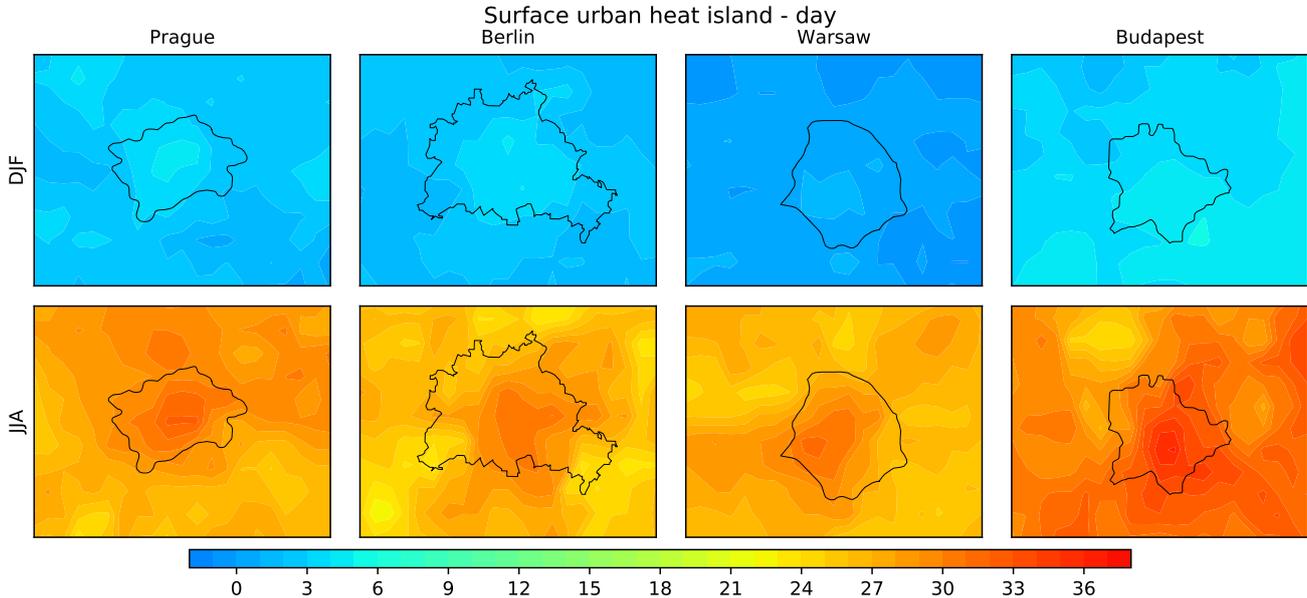
Figure 8. Satellite based daily winter and summer surface temperatures (°C) around Prague, Berlin, Warsaw and Budapest.

For further investigation of UMI components and to eliminate the effect of specific conditions occurring in different cities, we have chosen ten large European cities across the whole model domain (Tab. 4). For every city, we define the city centre together with three points in its surroundings in a distance of 30 km from the centre forming an equilateral triangle with one apex directed to the north. The bilinear interpolation is used to determine model values in the chosen locations. Values from three surroundings points are averaged to get one value for city centre and one value for vicinity for each city.

Firstly, different components of UMI are investigated in their diurnal cycles for every season. One sample t-test on 98 % of significance level was used to determine the statistical significance of non-zero difference between city centre and vicinity values averaged over all models and cities, separately for specific season and hour in day. The results for winter and summer seasons are shown in Fig. 10, with the red color meaning the statistical significance. UMI elements as air and surface temperature, boundary-layer, moisture flux are always significantly altered by urban surfaces, and with some exceptions also the wind speed, specific humidity and sub-grid scale precipitation. Cloud cover is influenced by cities in specific times of day, most continuously in summer afternoon and evening. The significance test did not reveal cross-model significant impact on wind direction and large-scale precipitation.

Alterations of temperature, wind speed and boundary-layer height in urban areas and their diurnal cycles are described in detail in many previous studies (e.g. Huszar et al., 2014; Huszár et al., 2018; Karlický et al., 2018; Huszar et al., 2020), therefore
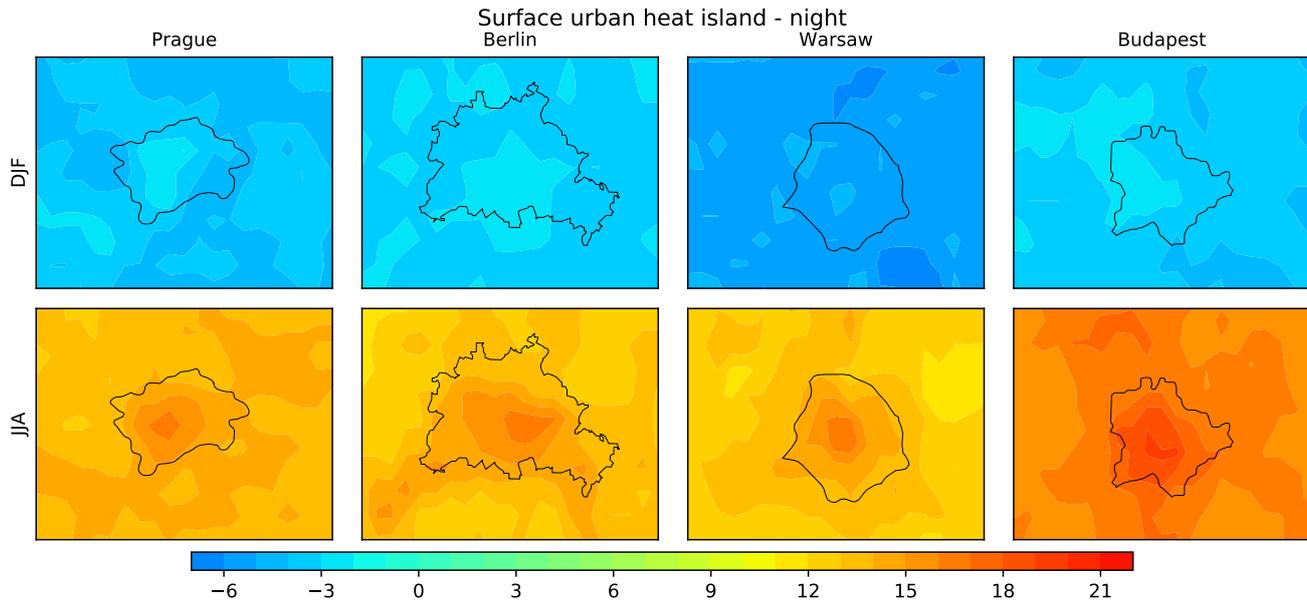
Figure 9. Same as Fig. 8 but for nighttime.

Table 4. Positions of city centres and vicinities for chosen cities.

	city centre		vic. 1 – 30 km N		vic. 2 – 30 km SWW		vic. 3 – 30 km SEE	
	lat	lon	lat	lon	lat	lon	lat	lon
Prague	50.075	14.44	50.345	14.44	49.94	14.076	49.94	14.804
Vienna	48.208	16.387	48.478	16.387	48.073	16.023	48.073	16.751
Budapest	47.5	19.076	47.77	19.076	47.365	18.712	47.365	19.44
Beograd	44.811	20.461	45.081	20.461	44.676	20.097	44.676	20.825
Zagreb	45.802	15.984	46.072	15.984	45.667	15.62	45.667	16.348
Warsaw	52.244	21.017	52.514	21.017	52.109	20.653	52.109	21.381
Berlin	52.521	13.408	52.791	13.408	52.386	13.044	52.386	13.772
Hamburg	53.594	9.986	53.864	9.986	53.459	9.622	53.459	10.35
Munich	48.147	11.567	48.417	11.567	48.012	11.203	48.012	11.931
Copenhagen	55.676	12.504	55.946	12.504	55.541	12.14	55.541	12.868
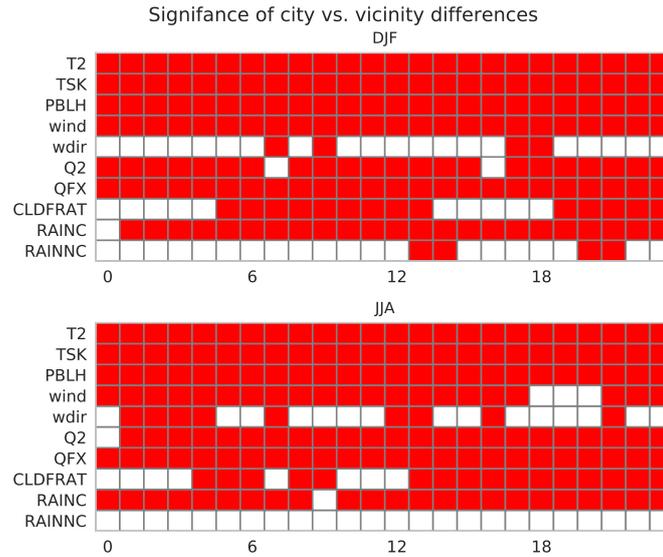
Figure 10. Significance of urban meteorology island components during the day. From above: 2-metre air temperature (T2), surface temperature (TSK), boundary-layer height (PBLH), wind speed (wind), wind direction (wdir), specific humidity (Q2), upward moisture flux (QFX), cloud cover (CLDFRAT), sub-grid scale precipitation (RAINC), large-scale (resolved) precipitation (RAINNC). Red color means statistical significance on the 98% level.

other components of the UMI will be discussed in more detail in the following paragraphs. In Fig. 11, differences between urban and rural cloud cover are presented. Statistically significant differences are detected mainly in summer afternoon and during evening, probably as a result of the enhanced convection, as the sub-grid scale precipitation is also increased in cities during this time (Fig. 12). In winter, a small but significant reduction of cloud cover is detected in models from morning to noon and during night.

Fig. 12 shows the alterations of sub-grid scale precipitation. The most distinct feature is the summer afternoon and evening increase, above 5 mm per season during some hours. However, a significant increase of sub-grid scale precipitation is detected nearly in the whole winter and summer diurnal cycle, but with much smaller magnitude: up to 1 mm in remaining summer hours and up to 0.3 mm in the winter diurnal cycle.

In terms of specific humidity, a statistically significant urban decrease prevails during summer and winter daytime (Fig. 13). The magnitude in summer (up to 1×10^{-3} kg kg^{-1}) is more pronounced compared to winter (up to 0.2×10^{-3} kg kg^{-1}). During nighttime, a slight humidity increases are detected (up to 0.1×10^{-3} kg kg^{-1}).

3.3 Impact of models and their parameterizations

Here we focus on the analysis of how different parameterization influences the resulting UMI component. Fig. 14 shows a spread of city-vicinity differences for different components of UMI, separately for every model simulation. In general,
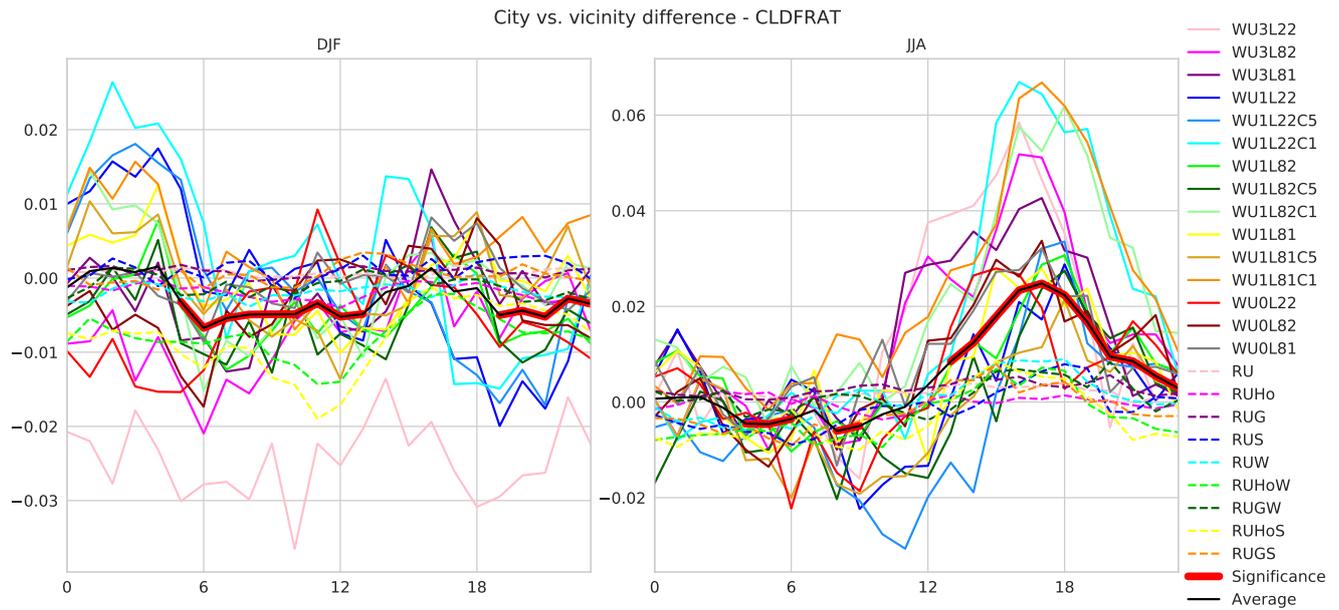
Figure 11. Diurnal cycle of city-vicinity differences in cloud cover (in fractions of one).
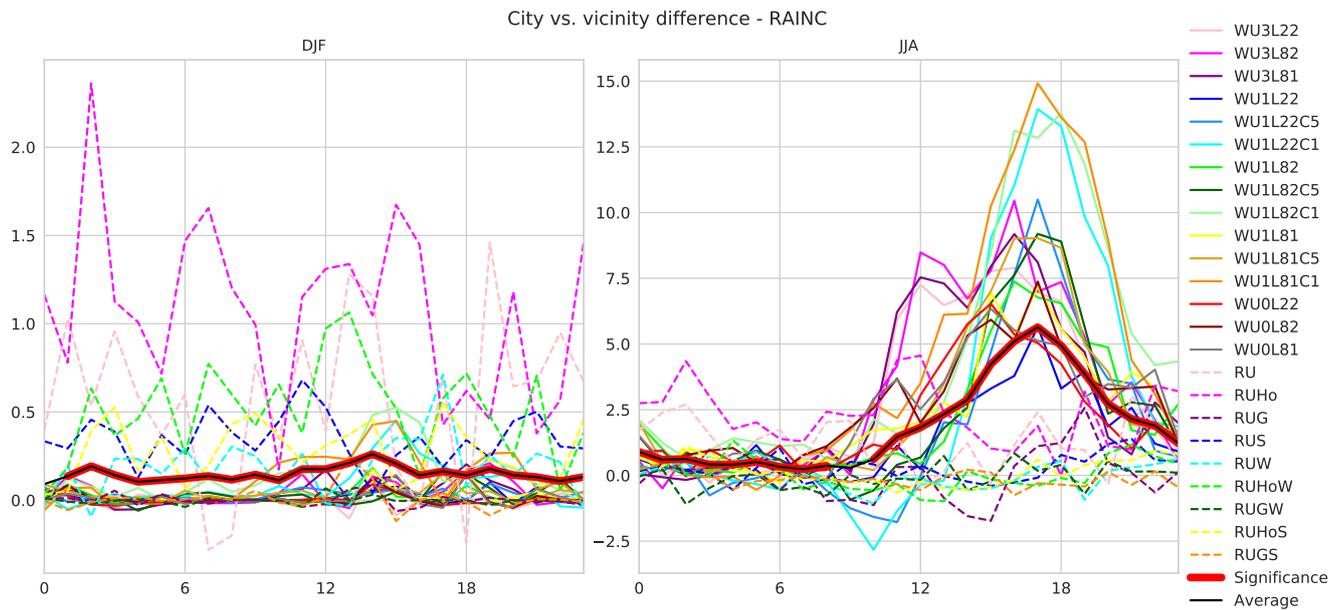

Figure 12. Diurnal cycle of city-vicinity differences for sub-grid scale precipitation (in mm, differences between seasonal sums).
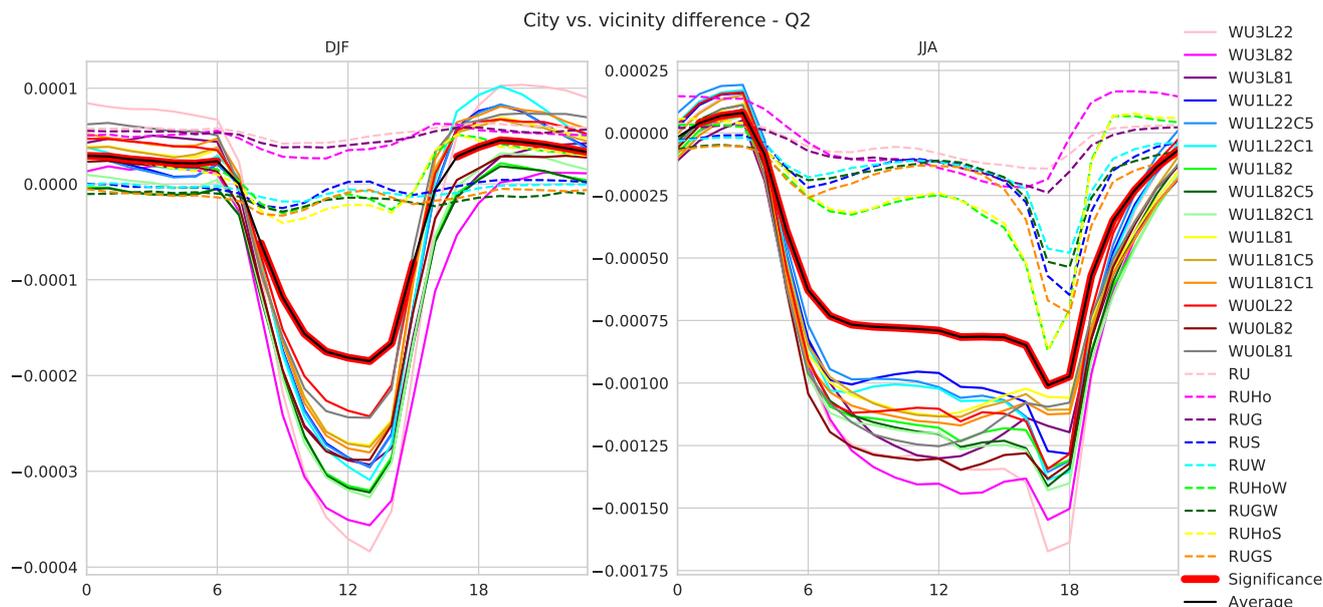
Figure 13. Diurnal cycle of city-vicinity differences for specific humidity (in $kg\,kg^{-1}$).

differences between WRF and RegCM results are large. In case of city-vicinity differences of seasonal daily temperature, in WRF simulations they are about 2 °C in summer and about 1–1.5 °C in winter, while simulations with the BEP+BEM urban parameterization make the difference even larger, up to 3 °C. On the other hand, in RegCM simulations, differences are much

205 lower – about 1 °C only. Similarly, in terms of specific humidity, city-vicinity differences in summer are also more pronounced in WRF simulations ($0.6–1.0 \times 10^{-3}$ $kg\,kg^{-1}$), while RegCM gives differences up to 0.3×10^{-3} $kg\,kg^{-1}$. In winter, the urban specific humidity reduction reaches 0.1×10^{-3} $kg\,kg^{-1}$.

Also the enhancement of PBL height in urban centers is greater in summer than in winter. In this season, the positive change is mainly between 200 and 300 metres, but in terms of RegCM model, only the simulations with the Holtslag PBL scheme

210 without Nogherotto MP reach such values. In winter, the difference is smaller: over cities, PBL is about 100 m higher than over vicinities and only the BEP+BEP urban model makes this difference higher. In terms of wind speed, the highest urban reductions are detected (independently on season) in WRF simulations with the BEP+BEP urban scheme. Simulations with the SLUCM urban scheme give the smallest reduction, but it also depends on the PBL parameterization. Combination of MYJ PBL and Eta SFL schemes makes the wind speed reduction about 1 $m\,s^{-1}$, combination of BouLac and Eta schemes

215 produces the reduction of about 0.5 $m\,s^{-1}$ and the combination of BouLac and MM5 schemes results in even a slight wind speed increase. In terms of WRF simulations with the bulk urban parameterization, it also depends on PBL and SFL schemes: in winter a reduction occurs but in summer the increase dominates. RegCM simulations exhibit urban wind speed reductions up to 1 $m\,s^{-1}$ and they are higher in winter than in summer.

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The impact of cities on cloud cover in summer is mostly positive, with a high dependency on model simulation. In winter, a cloud cover reduction dominates. In terms of sub-grid scale precipitation, WRF simulations produce significant summer increase in cities (mostly between 50 and 100 mm), but the winter urban modifications are negligible. In most RegCM simulations, the precipitation increase is much smaller in comparison to WRF simulations.

4 Discussion

The temperature underestimation of WRF simulations with the MYJ PBL and Eta SFL schemes (Fig. 3) in winter season are similar as detected by Karlický et al. (2018), summer temperatures are probably influenced by Tiedtke convection scheme (Tiedtke, 1989), which is not tested in this study, but making smaller summer overestimation of precipitation, related to higher summer temperatures. Also Zhong et al. (2017), who used the MYJ scheme of PBL in their study, show a slight temperature underestimation. BouLac PBL scheme seems to give more accurate temperature means and extremes. In terms of RegCM simulations, great temperature biases in simulations with the Nogherotto MP scheme imply a great overprediction of cloud cover, specific humidity and underestimation of downward shortwave radiation both in winter (Fig. 4) and summer (Fig. 5). This leads also to great precipitation biases and a slight temperature overprediction in winter. The remaining RegCM simulations show smaller temperature biases in summer, even less than Huszár et al. (2018), who used a very similar model setup. Huszar et al. (2020) used the model configuration corresponding to RU simulation (i.e. the Nogherotto MP, Tiedtke convection and UW PBL schemes), but it gives, in 9 km resolution, smaller biases: summer underestimation and winter overestimation of temperatures in range of 1–2 °C and precipitation overprediction by approximately 100 mm.

Focusing on cities, differences caused by distinct urban schemes appear. Winter positive bias of temperature in BEP+BEM simulations are well in-line with our previous study (Karlický et al., 2018). Huang et al. (2019) also described lower wind speeds and thereby more reduced bias in simulation with BEP+BEM in comparison to SLUCM. On the contrary, correlation of model and observed data of wind speed with values of the coefficient between 0.25 and 0.35 is substantially worse than in this study. In terms of temperatures, correlations are comparable.

Validation of model surface temperatures using MODIS satellite data (Fig. 6) results in similar features as the air temperature validation (Fig. 3), keeping in mind that MODIS monthly means are calculated only from days with clear-sky-condition. RegCM simulations with the Nogherotto MP scheme, by overestimating cloud cover during all seasons, show large negative biases (mainly in summer daytime). During nighttime, the positive cloud cover bias leads to temperature increase, mainly in winter. A general agreement of model surface temperatures with values derived from MODIS is shown also by Zhong et al. (2017) in annual averages.

Despite significant differences between specific simulations in terms of the UHI magnitude (Fig. 14), multi-model averaged UHI is still clearly visible around the selected big cities (Fig. 7). It is important to note that no adjustment on the same altitude was made so the results can be influenced partly by orography, which concerns mainly Prague and Budapest, due the relatively complicated terrain within and in the surroundings of these cities. The UHI magnitudes about 1 °C in winter and 1.5 °C in summer are in general agreement with Trusilova et al. (2016), who analysed the observation-based UHI for Berlin. However,
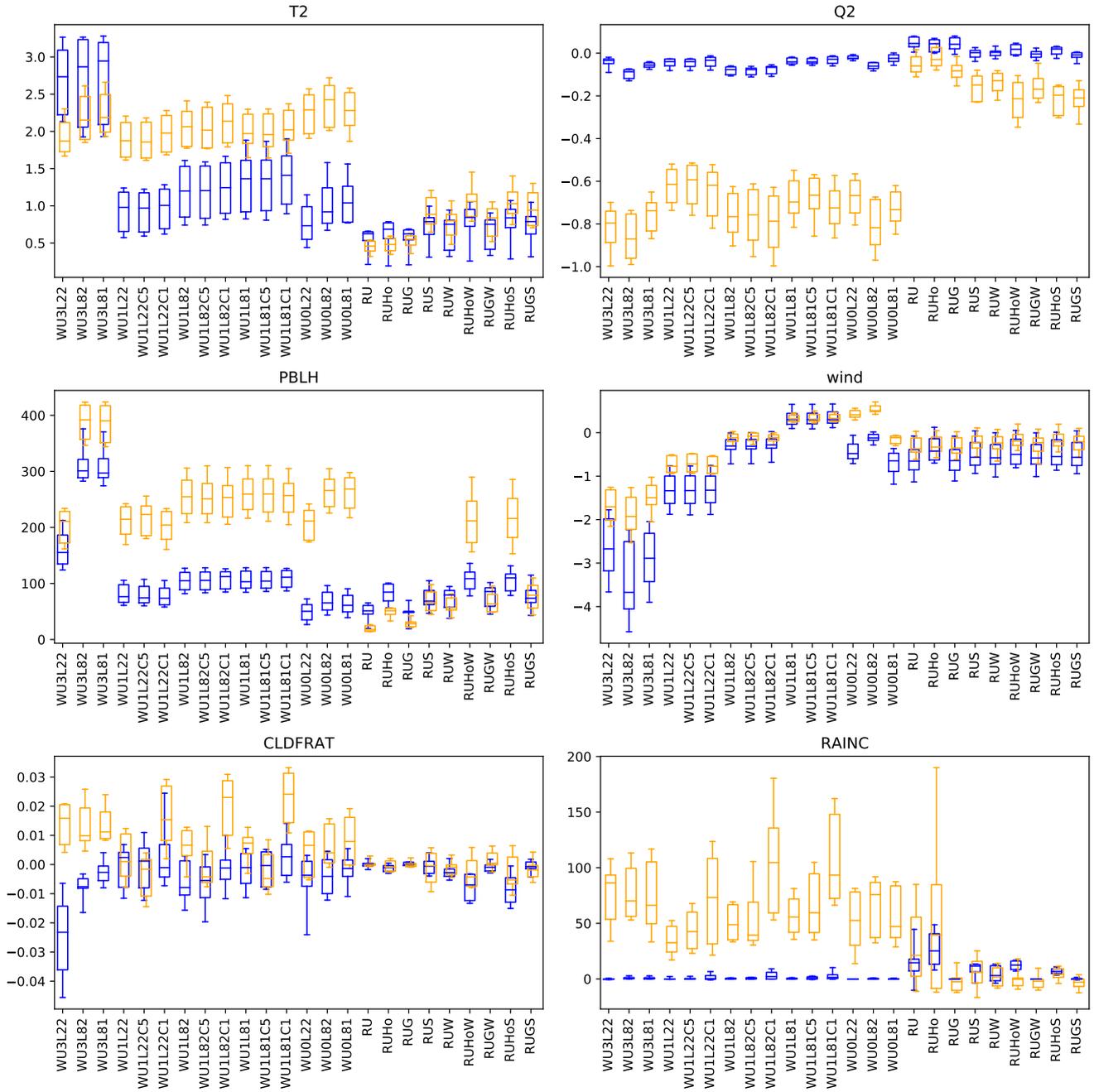
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Figure 14. Winter (blue) and summer (orange) differences between city and vicinity values of 2-metre air temperature (T2; in °C), specific humidity (Q2; 10^{-3} kg kg^{-1}), boundary-layer height (PBLH; m), wind speed (wind; m s^{-1}), cloud cover (CLDFRAT; fractions of one) and sub-grid scale precipitation (RAINC; mm in seasonal sums). Whiskers indicate 5th and 95th percentiles.

it depends on the location of stations: urban site 'Alexanderplatz' corresponds well with modelled summer UHI, but semi-urban sites 'Tempelhof' and 'Tegel' make the UHI less intensive. Langendijk et al. (2019), presenting results of a wide model ensemble, gives the annual UHI intensity for Berlin in 2015 almost 2 °C. Our previous study (Karlický et al., 2018), presenting observation-based data besides model results, gives smaller UHI intensities in Prague around 0.5 to 1 °C.

The satellite-based observations of surface temperature (Fig. 8 and Fig. 9) confirm the fact that the surface temperatures are significantly affected by urban surfaces too (Fig. 10). The intensities of SUHI around Budapest (Fig. 8 and Fig. 9) are less than given by Göndöcs et al. (2017), who also investigated daily and nocturnal SUHI by MODIS around this city and reached for the city centre and one-week period in summer, 6 °C and 4 °C for daytime and nighttime, respectively. Zhong et al. (2017), who presented an annual mean of surface temperatures taken from MODIS for Yangtze River Delta region of China, give the intensity of SUHI above 2 °C, which is comparable to our results, if we consider daily and nocturnal values in winter and summer. Satellites sampled the surface temperature only during clear-sky conditions, which means that the SUHI deduced from these measurements is somewhat stronger than the average for the whole season, given the fact that SUHI is more pronounced during sunny days with higher solar input.

Significance of urban-induced alterations of urban canopy air temperature, surface temperature, PBL height is well documented in many previous simulation and observation-based studies (e.g. Huszar et al., 2014; Trusilova et al., 2016; Göndöcs et al., 2017; Karlický et al., 2018; Huszár et al., 2018; Huszar et al., 2018, 2020). Also the moisture flux decrease in cities is a well-known phenomenon as shown already by Oke (1987), or more recently by Theeuwes et al. (2019) from latent heat flux comparison. In terms of wind speed changes in urban areas, no statistically significant differences occur during summer evening hours, when the cross-model average is nearly zero. This is probably caused by the combination of UHI, enhanced turbulence, convection and mixing in a deeper urban boundary-layer, as described by Droste et al. (2018), who concluded that under special conditions and during certain daytimes, the urban wind speed can be even higher.

Impact of cities on the cloud fraction (Fig. 11) is well documented also by Theeuwes et al. (2019), who investigated cloud cover and its differences in Paris and London and their surroundings during the warm season. They also found a cloud cover urban increase during afternoon and evening, despite drier atmosphere above cities that leads to a cloud base located higher by approximately 250 m. Higher temperatures in cities probably result in partial dissolution of non-precipitation stratiform clouds and fog in winter, with statistically significant cloud cover reduction during morning hours, because urban precipitation is not reduced (but rather increased, Fig. 12). The reduction of low-level clouds and fog was also found by Yan et al. (2020). In summer, higher temperatures lead to enhanced convection and more frequent occurrence of convective clouds, which leads to increase of sub-grid scale precipitation.

The summer afternoon and evening sub-grid scale precipitation enhancement in cities is also documented by Manola et al. (2020), who analysed observed precipitation features in Amsterdam and its surroundings. Differences occur in summer morning, when they found an increase of the same magnitude as in afternoon. Secondly, they found that the impact of cities on precipitation is higher in winter in relative numbers, which contrasts to our results (except one RegCM experiment). It can be partly explained by choice of cities for investigation (Table 4), that are located in central Europe with greater distance to

oceans and, thus, less maritime climate compared to Amsterdam. Zhu et al. (2019), who analysed the impact of urban areas on precipitation in the Beijing area, also found its positive, but only in the time range of 10–21 hour, which is close to our results.

In terms of impact of cities on specific humidity, Langendijk et al. (2019) gives for the annual difference between Berlin and its surroundings about 0.7×10^{-3} kg kg^{-1}, which corresponds to our summer values (Fig. 13) and rather to WRF simulations (Fig. 14). Winter differences are much less, even in WRF simulations. During nighttime, a slight humidity increase is visible (Fig. 13), caused probably by reduced occurrence of dew in cities connected to smaller humidity losses by condensation on surfaces, given that latent heat in cities is still positive during nights, in contrast to surrounding areas (Theeuwes et al., 2019). The daytime urban humidity decrease can be explained by less water availability, enhanced convection and higher vertical turbulent mixing (higher PBL) over cities.

The differences between WRF and RegCM models in terms of intensities of specific UMI elements are largely caused by the fact that WRF uses dominant land-use (Fig. 1), while RegCM model considers fractional land-use (Fig. 2), so the urban effects in RegCM simulations do not have to be so intensive in city centres and, on the other hand, they are non-zero in their surroundings, where non-zero urban fractions occur, which leads to smaller differences between cities and and their rural vicinities. In other words, the modelled UMI is smoother. This feature is responsible for less UHI intensities in RegCM simulation (Fig. 14). In terms of WRF, higher winter urban temperatures in simulations with the BEP+BEM urban scheme are consistent with results of previous studies (Liao et al., 2014; Karlický et al., 2018) and are probably caused by higher amounts of anthropogenic heat, internally computed within the scheme. In simulations with the bulk urban scheme, no anthropogenic heat is considered and therefore the winter UHI is suppressed.

As expected, differences between urban and rural PBL heights are sensitive, besides on model and urban scheme, also on the chosen parameterization of boundary-layer processes. In WRF simulations, the MYJ PBL scheme produces a smaller urban PBL increase compared to the BouLac PBL scheme. Halenka et al. (2019), also using WRF and RegCM to investigating urban effects over Prague, reported urban PBL increases only about 50 m in winter and 100 m in summer in terms of WRF and about 100 m in terms of RegCM (using the Holtslag PBL scheme without Nogherotto MP scheme). In terms of the urbanization induced wind speed changes, again, parameterization of urban processes and PBL are the main influencing factors. Also Halenka et al. (2019) gave the highest wind speed reduction in BEP+BEM simulation, in range 1.5–2 m s^{-1}, but in our case of more cities (Fig. 14), the reduction is much higher. In terms of SLUCM simulations, the reduction is similar, about 1 m s^{-1}, similarly in RegCM simulations (about 0.5 m s^{-1} reduction). The wind speed urban increase in bulk simulation is visible only in the summer season in Halenka et al. (2019). The wind speed reduction is mostly less in summer, when higher urban roughness can be compensated by enhanced turbulence, convection and mixing (Droste et al., 2018), i.e. many counteracting effects play a simultaneous role and each model configuration prefers only a subset of these effects.

The impact of cities on cloud cover is clearly influenced by the convection parameterization: Kain-Fritsch scheme makes the summer urban cloud cover increase stronger, despite the fact that this setup does not give the highest precipitation in general. This is very similar to the summer urban increase of sub-grid scale precipitation. The great increases in RegCM simulations with the Nogherotto MP scheme can be explained by overall precipitation bias, where increased urban convection leads to higher difference. In all simulations, it seems that urban changes of sub-grid scale precipitation are largely city-dependent,

in case of RegCM this concerns even the sign of the change. It indicates that some other climate elements such as seasonal total precipitation climatology may influence the urban increase of sub-grid scale precipitation and more research has to be conducted in this regard.

5 Conclusions

325 The study presented results of a model ensemble of 24 simulations using WRF and RegCM models performed over a European domain with 9 km horizontal resolution and covering a two year period. Such a great ensemble enables the robust investigation of the impact of urban surfaces on overall climate and weather in cities, because model uncertainties given by specific model setups were eliminated in the whole ensemble. Urban-induced changes, manifesting in various meteorological variables, were generalised as one concept called urban meteorology island (UMI), where urban perturbations of different meteorological

330 quantities were regarded as components of the overall UMI.

The results of the study showed that this concept is justified and our approach is meaningful, because almost all investigated meteorological variables are significantly altered in cities with respect to their rural counterparts, independently to the chosen city and the model setup. However, quantitative or even (to some extent) qualitative differences were detected between models.

The main conclusions of the presented study are:

335 • Validation showed large differences between individual model simulations given by different parameterizations of the different physical processes driving urban meteorology. In RegCM, the microphysics parameterization had the greatest impact on temperature and precipitation biases. The impact of other parameterizations is smaller; Tiedtke convective scheme overpredicts precipitation less than Grell scheme and the Holtslag PBL scheme slightly improves temperature. In WRF, temperature biases are smaller, especially in simulations with BouLac PBL scheme, and significant overestimation

340 is encountered mainly for summer precipitation.

• For ECAD stations (often in cities), the type of urban canopy schemes turned out to be the most important factor determining the model's accuracy. The BEP+BEM urban scheme gives winter temperature minima about 2 °C higher than observed values and reduces positive wind speed bias by 0.5–1 $\mathrm{m\,s^{-1}}$. Most of the model simulations correlate well with ECAD observations, with the correlation coefficient about 0.8 or higher, with exception of cloud cover and relative

345 humidity in winter.

• In general, the UMI components are more pronounced in WRF which used dominant land-use compared to RegCM fractional land-use approach, pointing out the importance of the land-use model representation.

• An increase of cloud cover in cities is modelled, mainly for summer afternoons. This is connected with a summer afternoon urban increase of sub-grid scale precipitation (by 5 mm). These changes are probably caused by an enhanced

350 convection over urban areas, given by higher near surface temperatures while specific humidity is significantly lower in cities.

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- Impact of urban and other parameterization: the BEP+BEM urban model increases the winter UHI by more than 2 °C, and leads to highest reductions of urban wind speeds. In terms of PBL height and its urban increase in summer, the MYJ PBL scheme in WRF simulations makes this UMI element about 50 m smaller, while Holtslag PBL scheme in RegCM increases it to the WRF level. The choice of PBL scheme further influences the wind speed reduction in cities. Modifications of cloud cover and sub-grid scale precipitation in cities are influenced mainly by the parameterization of convection and here, the Kain-Fritsch scheme results in the most pronounced UMI, mainly during summer season.

- Besides results based on model simulations, satellite measurements of surface temperature show significant increase in urban areas too, with the magnitude 3–4 °C in summer and about 2 °C in winter, during clear-sky days.

Our study showed the great importance of multi-model approach when describing the urban meteorological phenomenon, as large differences exist between models in their way of resolving different city and regional scale physical processes. Future research on urban atmospheric processes should thus be based on model/physical ensembles rather than a single model experiments.

Code and data availability. Source codes of the WRF and RegCM models are publicly available. Resulting data from simulation performed (stored in the Charles University data storage facilities making about 30 TB) can be obtained upon request to authors.

Author contributions. PH designed the main scientific idea and organised the project team, in collaboration with TH. MB prepared the input and validation data; JK and PH performed the model simulation; JK analysed the resulting data, with a contribution of TN, FŠ and JĎ. JK wrote the text and PH revised it.

Competing interests. The authors declare that they have no conflict of interest.

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