



Contribution to AEVUS-2 PT by RHM & Moscow State University: Key results and recent updates

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Set up of the recent simulations

Initial and boundary conditions:

Continuous data set, constructed from ICON forecasts for 3 and 6 hours:

Forecast from 0 UTC

| | | | | | |
|---|---|---|---|----|-----|
| 0 | 3 | 6 | 9 | 12 | ... |
|---|---|---|---|----|-----|

Forecast from 6 UTC

| | | | | | |
|---|---|---|---|----|-----|
| 0 | 3 | 6 | 9 | 12 | ... |
|---|---|---|---|----|-----|

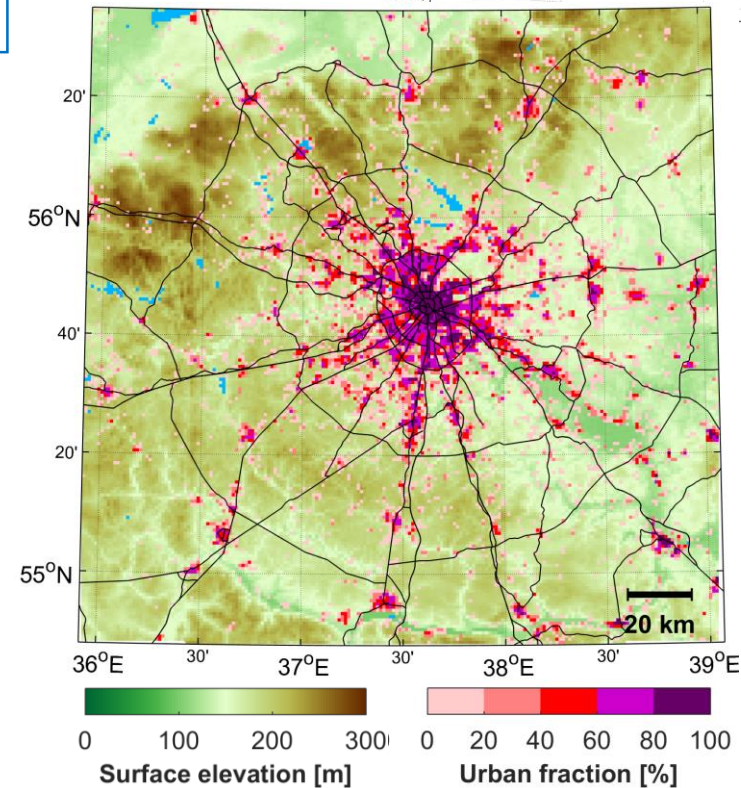
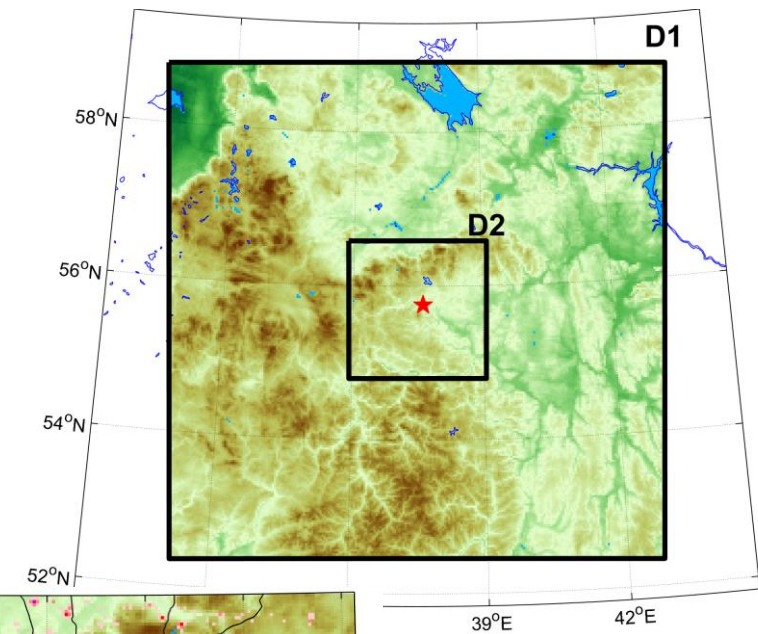
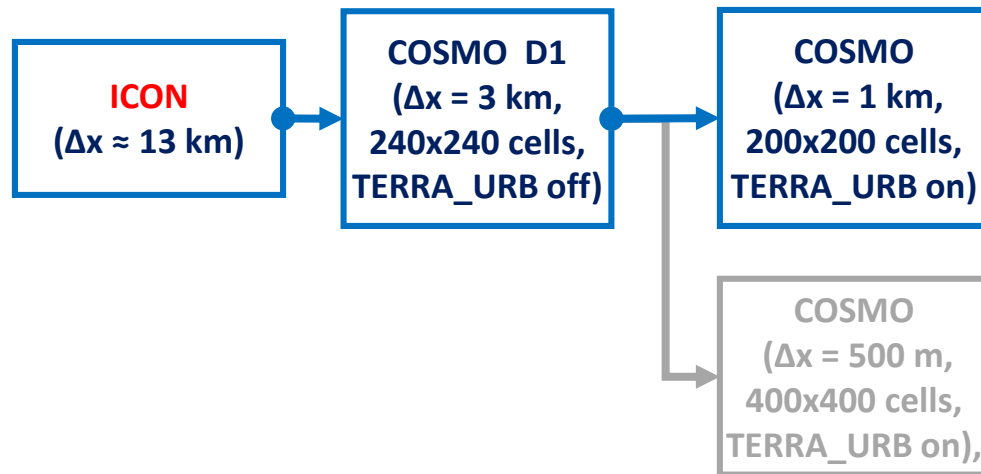
Forecast from 12 UTC

| | | | | | |
|---|---|---|---|----|-----|
| 0 | 3 | 6 | 9 | 12 | ... |
|---|---|---|---|----|-----|

Continuous data:

| | | | | | | |
|---|---|----|----|----|----|-----|
| 6 | 9 | 12 | 15 | 18 | 21 | ... |
|---|---|----|----|----|----|-----|

Scheme of dynamical downscaling:



Case studies:

- 1) 5-20 August 2017
- 2) 1-15 June 2019

Set up of the recent simulations

- ✓ TERRA_URB is always on
- ✓ Namelist settings generally similar to ARPAP's one, with exception for **itype_aerosol** (2 instead 1), **hcorr_*** parameters (model defaults are used), **llake** (TRUE instead FALSE) and some parameters for new turbulence

For common AEVUS paper?

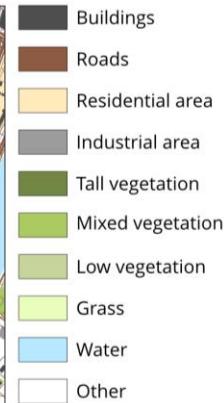
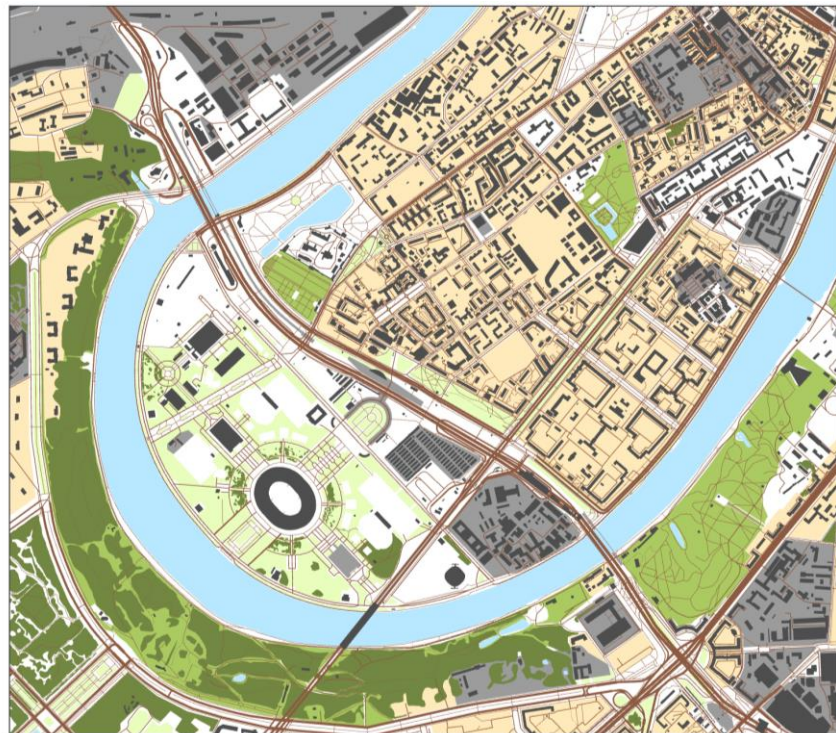
| Simulation number | Turbulence | itype_canopy | itype_vdif | Urban canopy parameters | Urban thermal parameters |
|-------------------|------------|--------------|------------|--|--|
| AEV1 | Old | 1 | -1 | Detailed "local" data set based on Openstreetmap, CGLC & Sentinel data | Defaults from (Wouters et al., 2016) |
| AEV2 | Old | 2 | -1 | | |
| AEV3 | New* | 1 | -1 | | |
| AEV4 | New* | 2 | -1 | | |
| AEV5 | New* | 2 | 1 | | |
| AEV5a | New* | 2 | 1 | LCZ-based data from Mathias | Modified values based on literature review |
| AEV5b | New* | 2 | 1 | | |

*Settings for "new turbulence" include **pat_len**=100 (model default, instead of 750 from User Guide), **c_soil** = 1 (model default, instead of 1.75/1.25 from User Guide)

Urban canopy parameters: custom data set

Based on combined use of different data sets

- Built up fraction area from *Copernicus Global Land Cover* with 100 m resolution
- Data on buildings and roads from *OpenStreetMap*
- Data on vegetation derived from *Sentinel-2 satellite images* with 10 m resolution



Trees and grass in 10 m resolution



Copernicus Global Land Service
Providing bio-geophysical products of global land surface

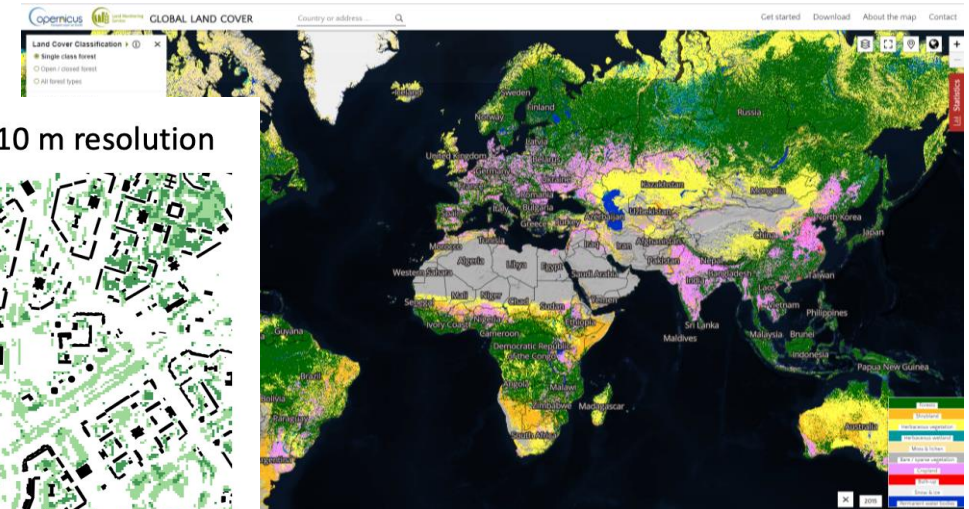


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Release of Global 100m Land Cover maps for 2015

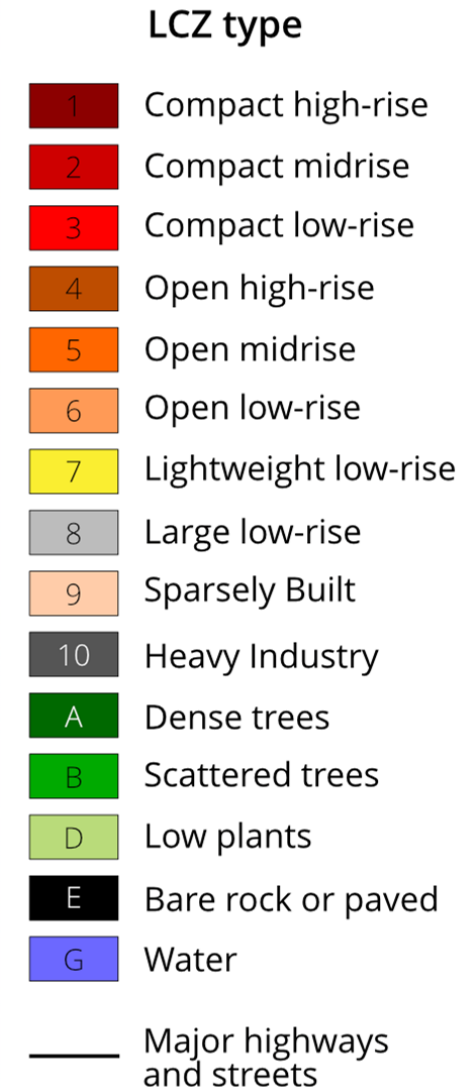
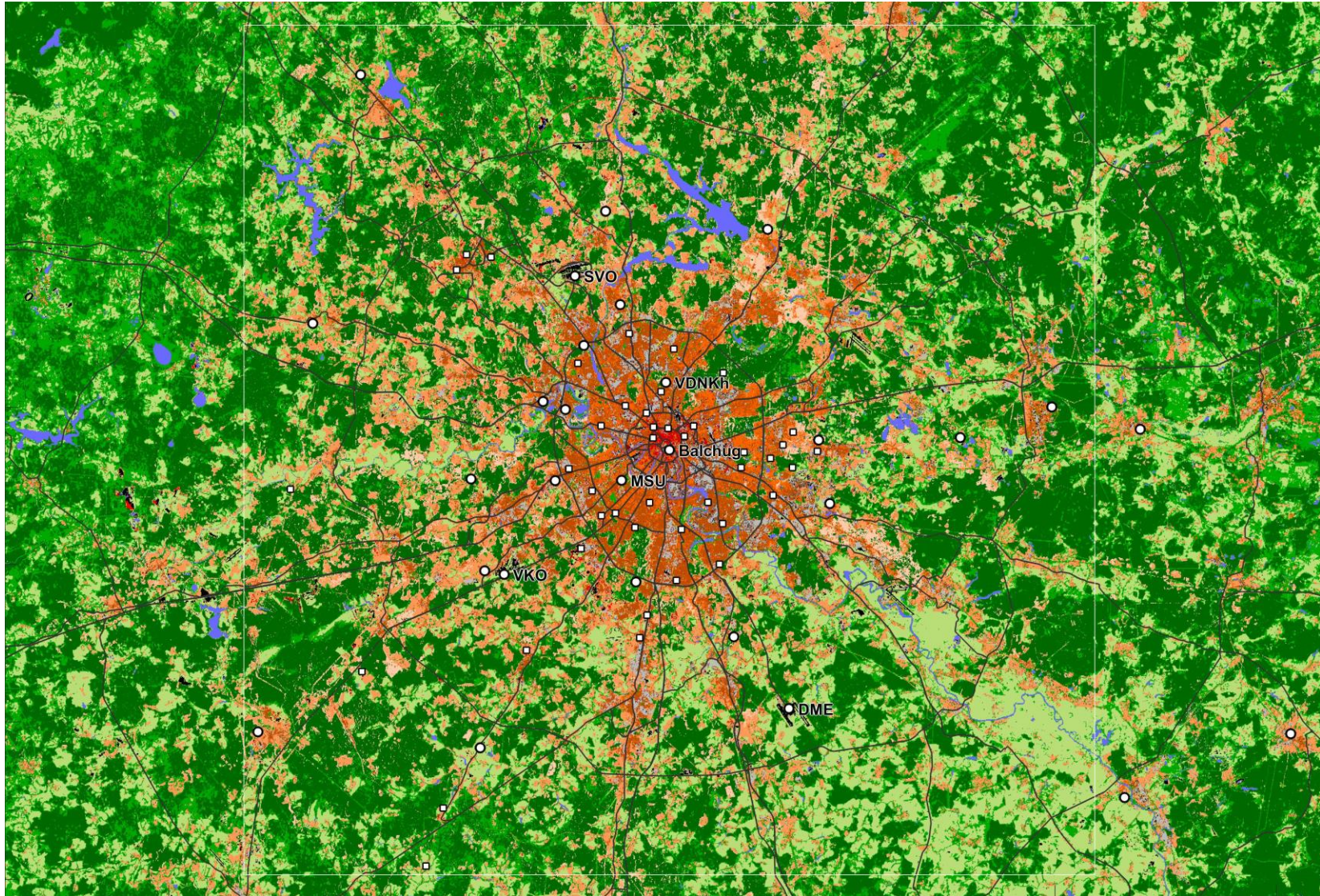
Today, at the occasion of ESA's biggest Earth observation conference, the 'Living Planet Symposium 2019' (Milan, Italy), the Global Land Service team is thrilled to **release** a new set of **Global Land Cover** layers, with an **overall 80% accuracy**:

- a complete, **discrete classification with 23 classes**
- **fractional cover layers** for the **ten** base land cover classes: forest, shrub, grass, moss & lichen, bare & sparse vegetation, cropland, built-up / urban, snow & ice, seasonal & permanent inland water bodies.
- a **forest type layer** offering twelve types of forest
- **quality indicators** for input data (data density indicator), for the discrete map (probability) and for six of the fractional cover layers.



(Samsonov, Varentsov, 2020, submitted)

Urban canopy parameters: LCZ-based data set



Open high-rise



Sparsely built

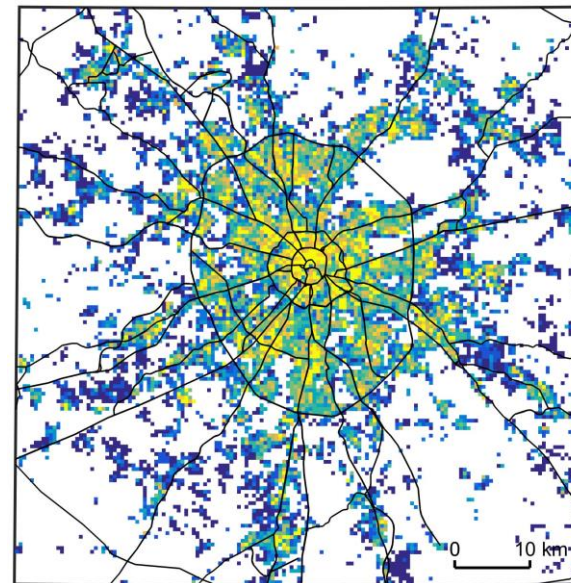
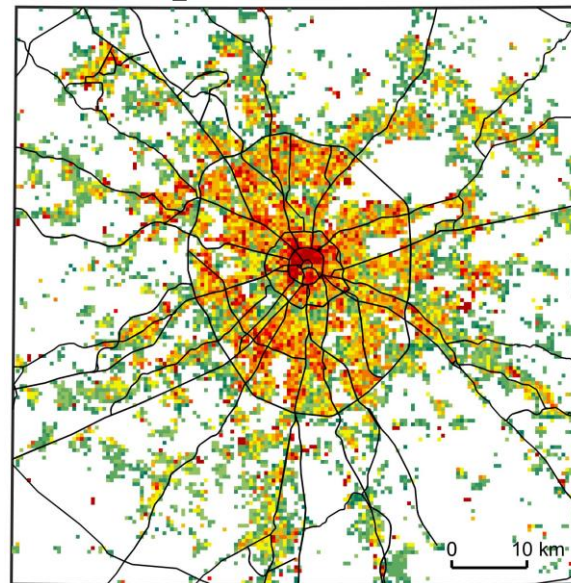
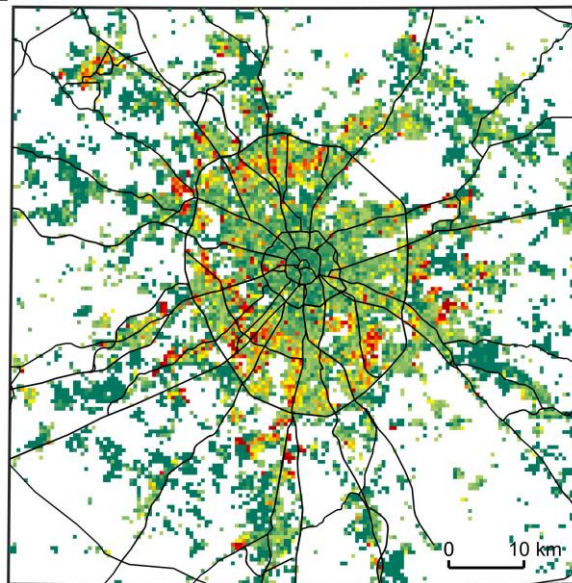
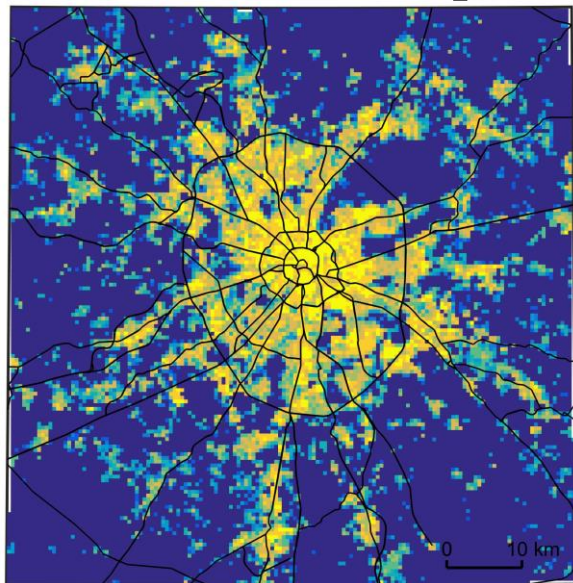


Low plants

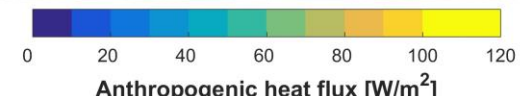
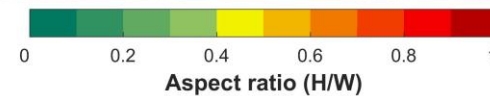
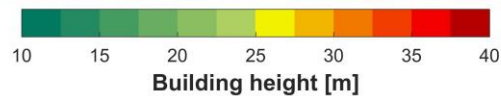
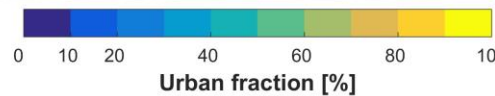
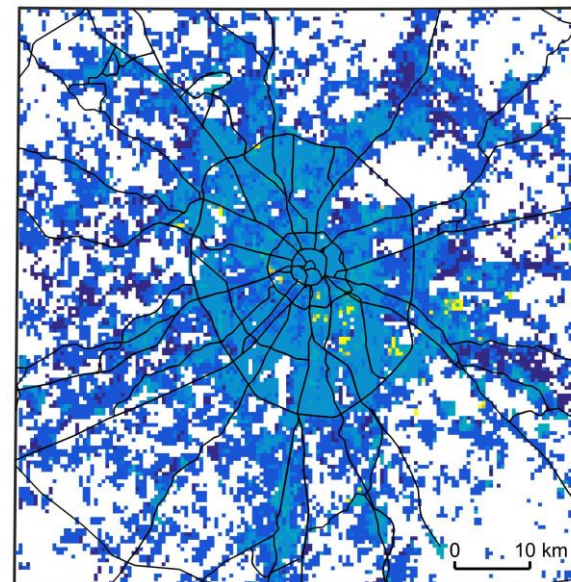
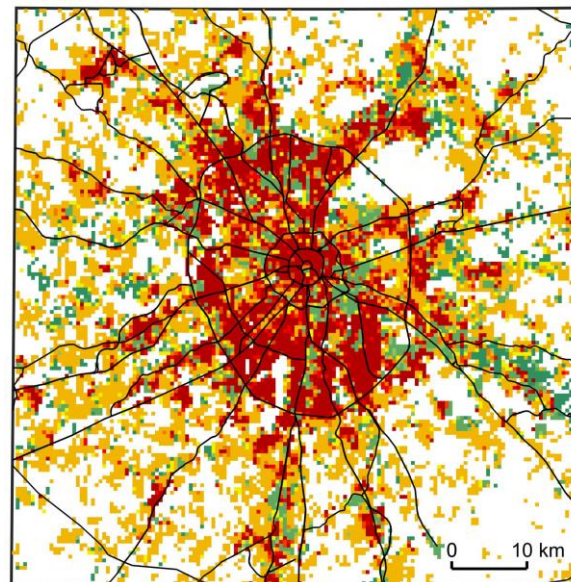
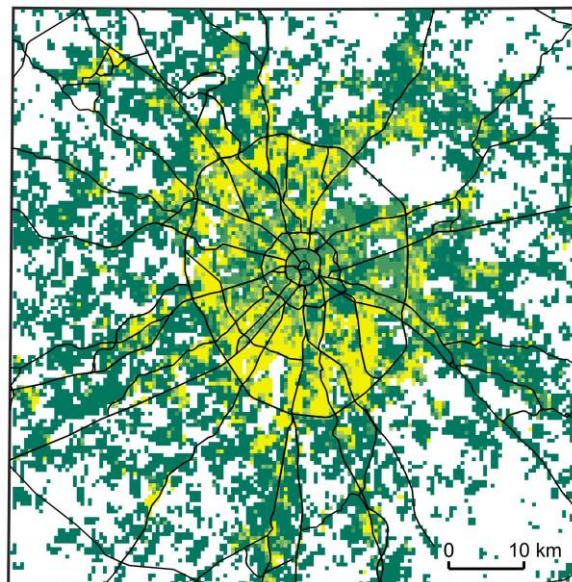
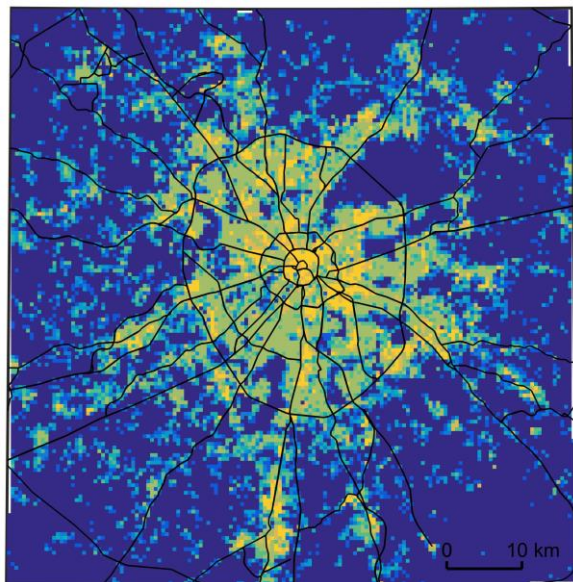
LCZ map for Moscow region from (Samsonov, Trigub, 2017) was recently re-classified and extended for a wider area by Matthias Demuzere (Ruhr University Bochum)

Urban canopy parameters: comparison

Custom data set



LCZ-based UCPS

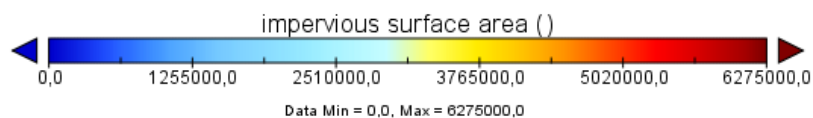
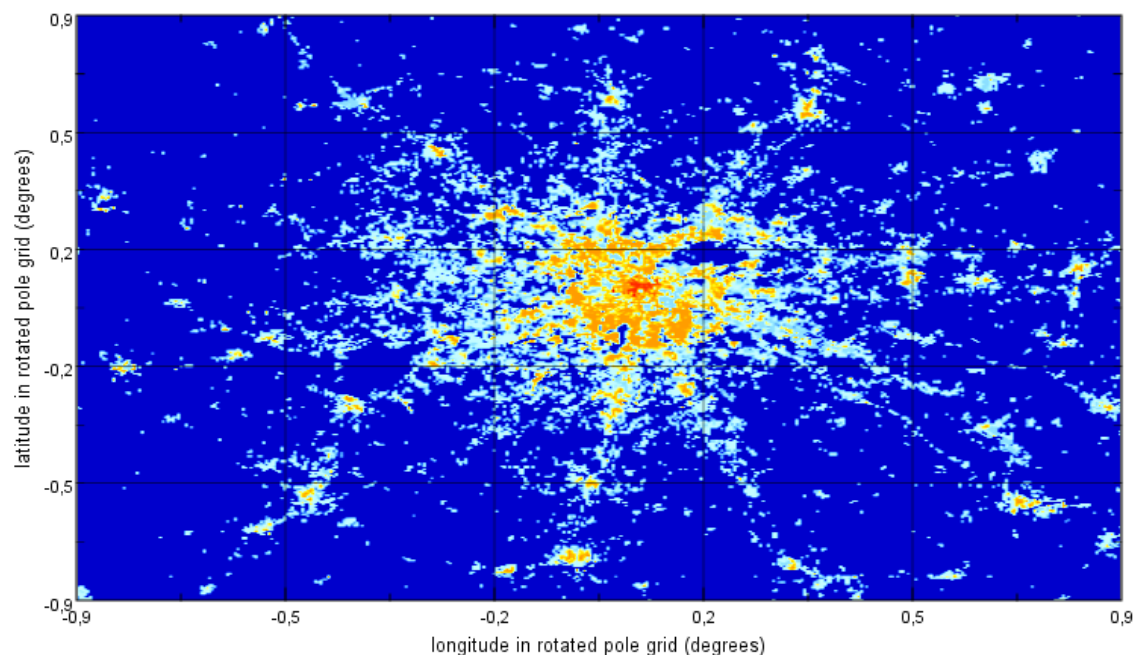


What about urban thermal parameters?

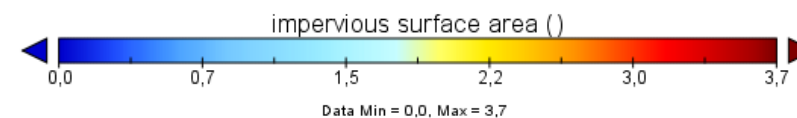
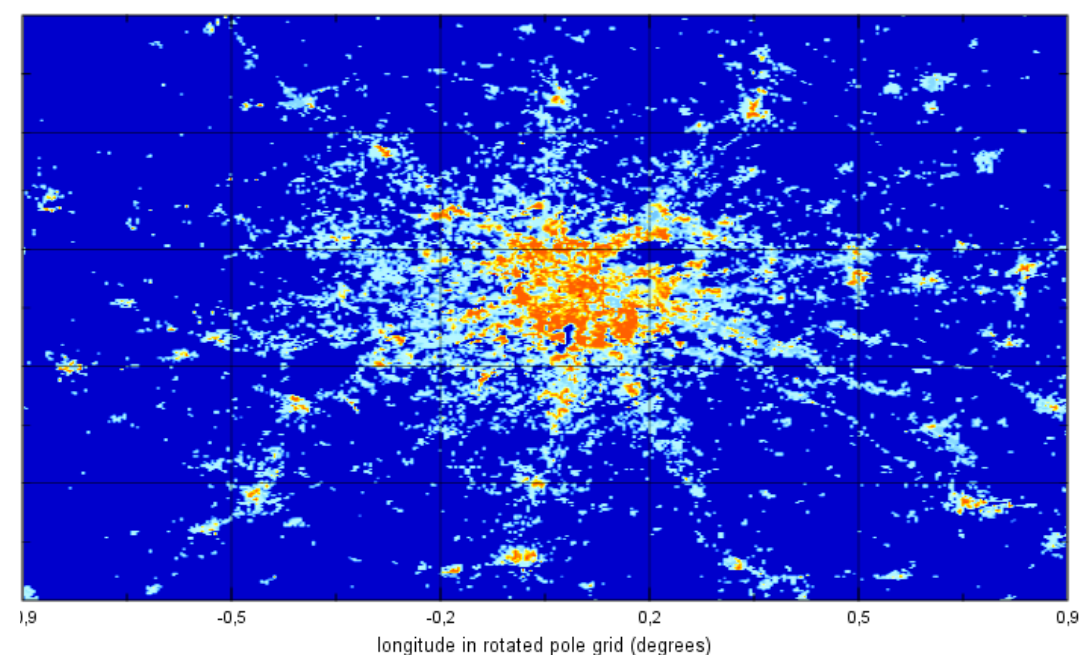
LCZ-based values looks strange, additional check is needed, so not used yet

| | default values from (Wouters et al., 2016) | Probably more reasonable values obtained based on literature review for AEV5a run |
|-----------|--|---|
| curb_hcap | $1.25 \cdot 10^6$ | $1.8 \cdot 10^6$ |
| curb_hcon | 0.77 | 0.77 |
| curb_salb | 0.1 | 0.2 |
| curb_tabl | 0.14 | 0.14 |

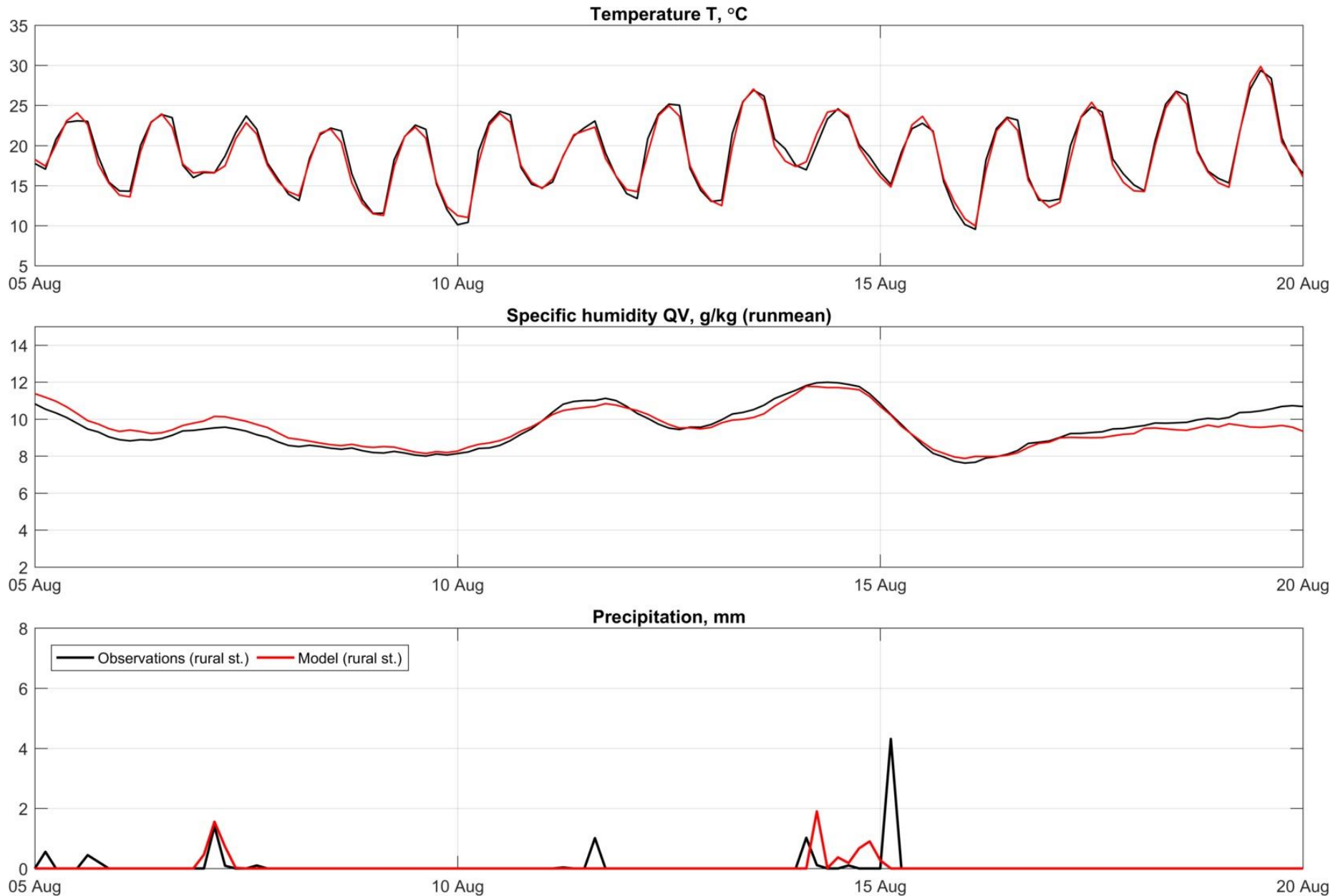
URB_HCAP



URB_HCON

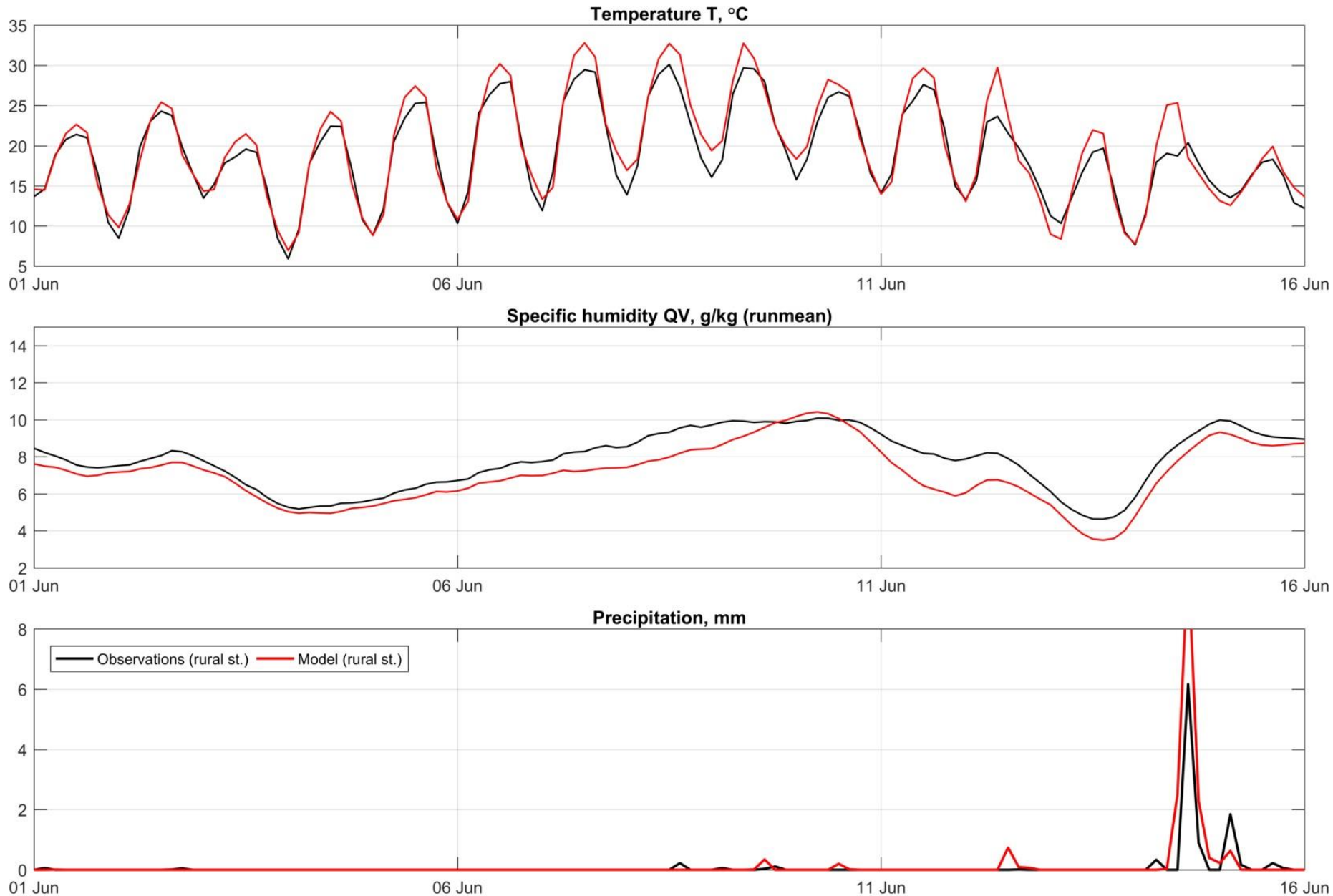


Basic verification: case 1 (05.08.2017-20.08.2017)



**Results for D1 domain
(no TERRA_URB),
mean values for 9 rural
stations,
AEV5 simulation**

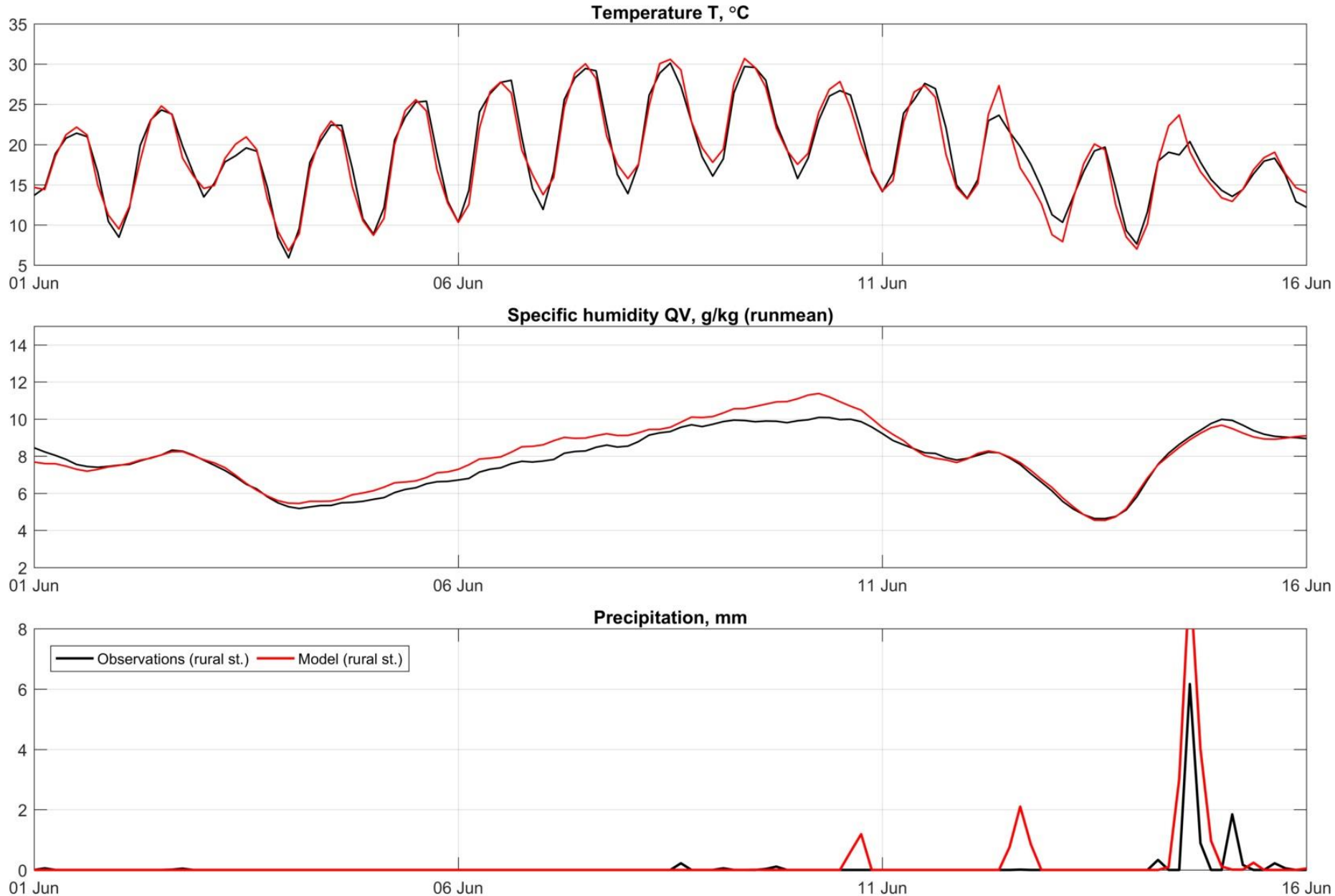
Basic verification: case 2 (01.06.2019-15.06.2019)



Results for D1 domain
(no TERRA_URB),
mean values for 9 rural
stations,
AEV5 simulation

**Problem: warn & dry bias
over whole domain**

Basic verification: case 2 (01.06.2019-15.06.2019)

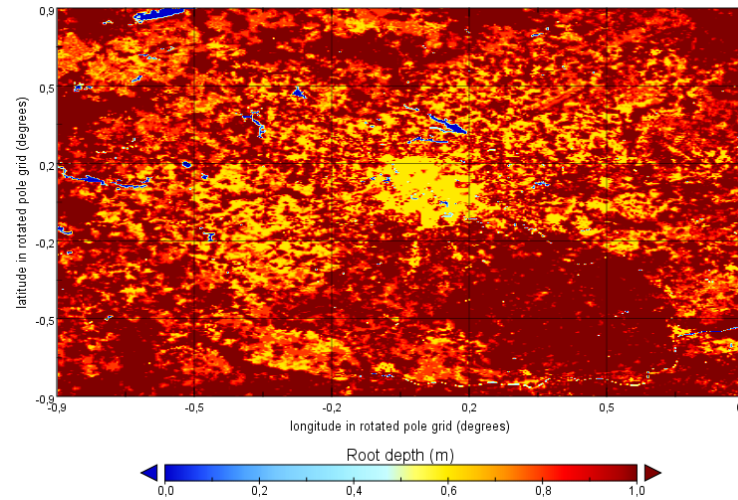
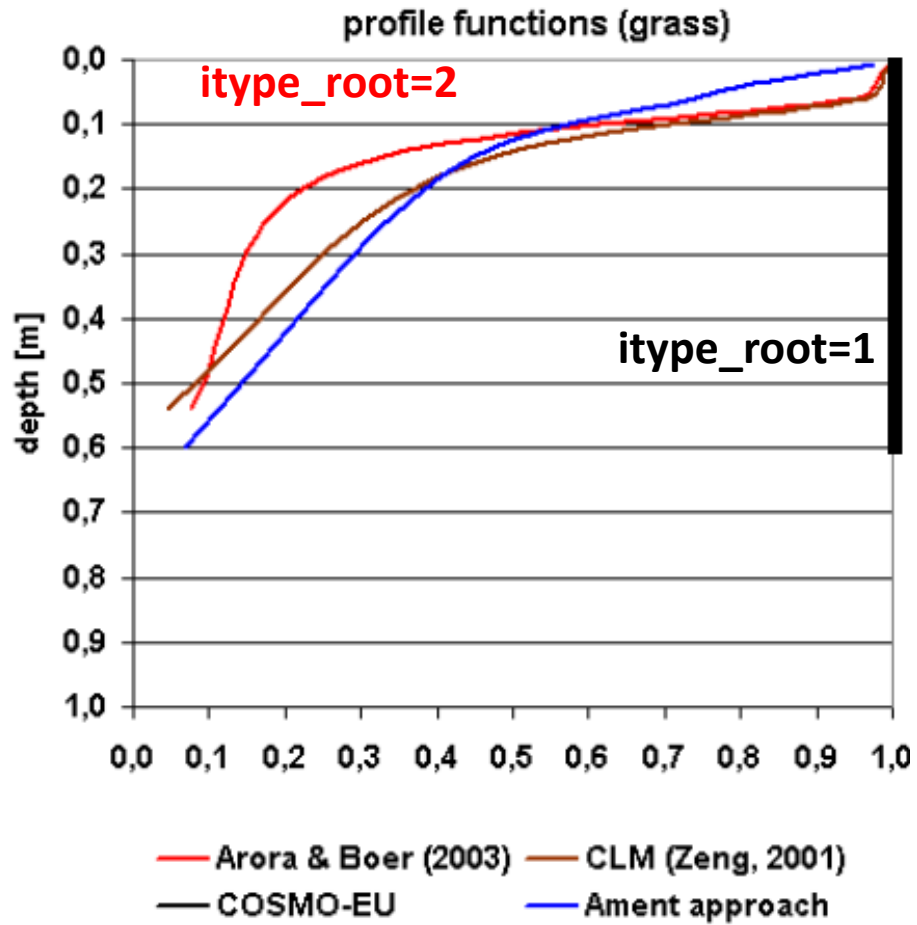


Results for D1 domain
(no TERRA_URB),
mean values for 9 rural
stations,
AEV5 simulation

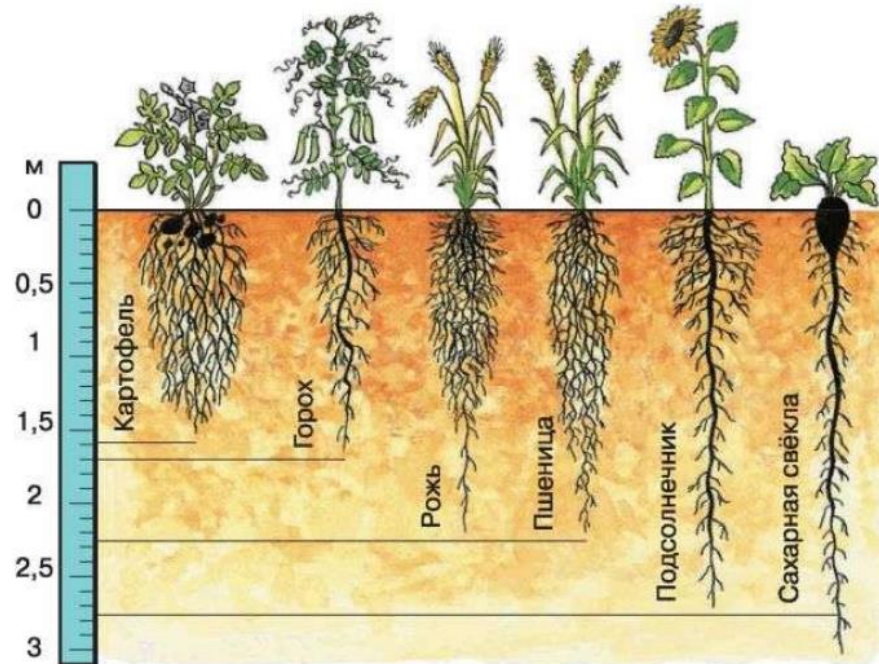
**Problem: warm & dry bias
over whole domain**

**Solution: increased
rooting depth by 2.5
times (crootdp = 2.5)
in all simulations for this
case**

Why rooting depth?



Rooting depth from
Extpar:
No higher than 1 m

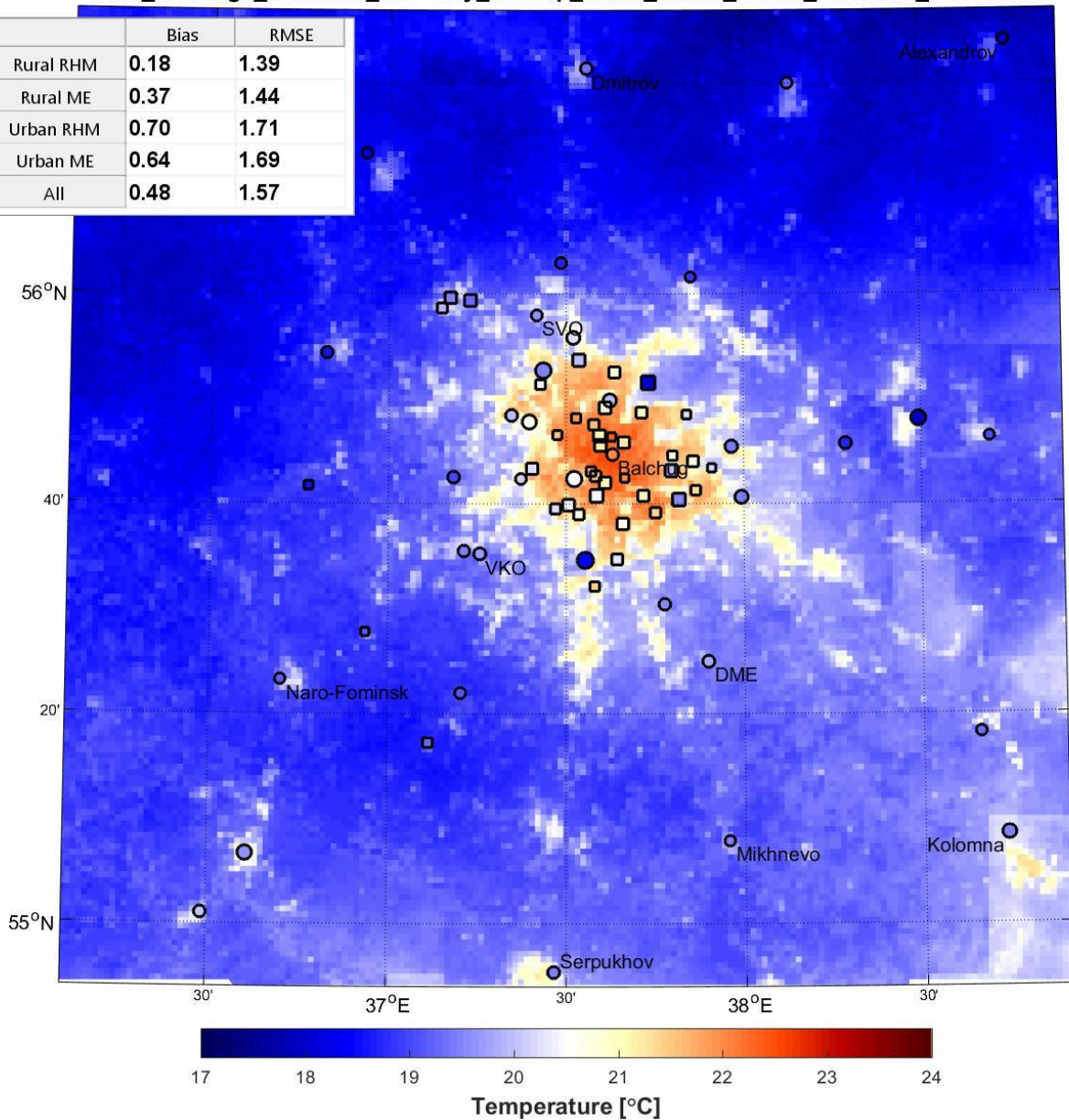


Rooting depth
in real world:
**1.5 m even for potatoes
(and what about trees?)**

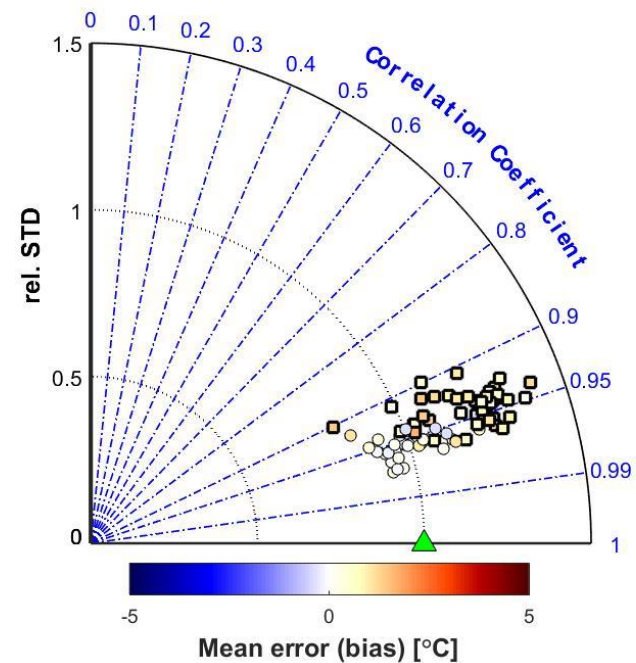
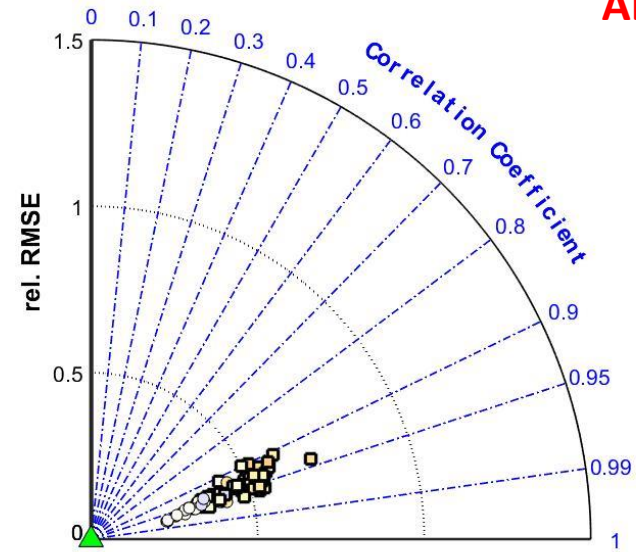
Verification for urban areas: example for case 1

MSK_0.009bg3_osmurb2_v2c Cray_v505up_AEV1_oldtur_noskc_osmurb2_v2c L3n9u

| | Bias | RMSE |
|-----------|------|------|
| Rural RHM | 0.18 | 1.39 |
| Rural ME | 0.37 | 1.44 |
| Urban RHM | 0.70 | 1.71 |
| Urban ME | 0.64 | 1.69 |
| All | 0.48 | 1.57 |



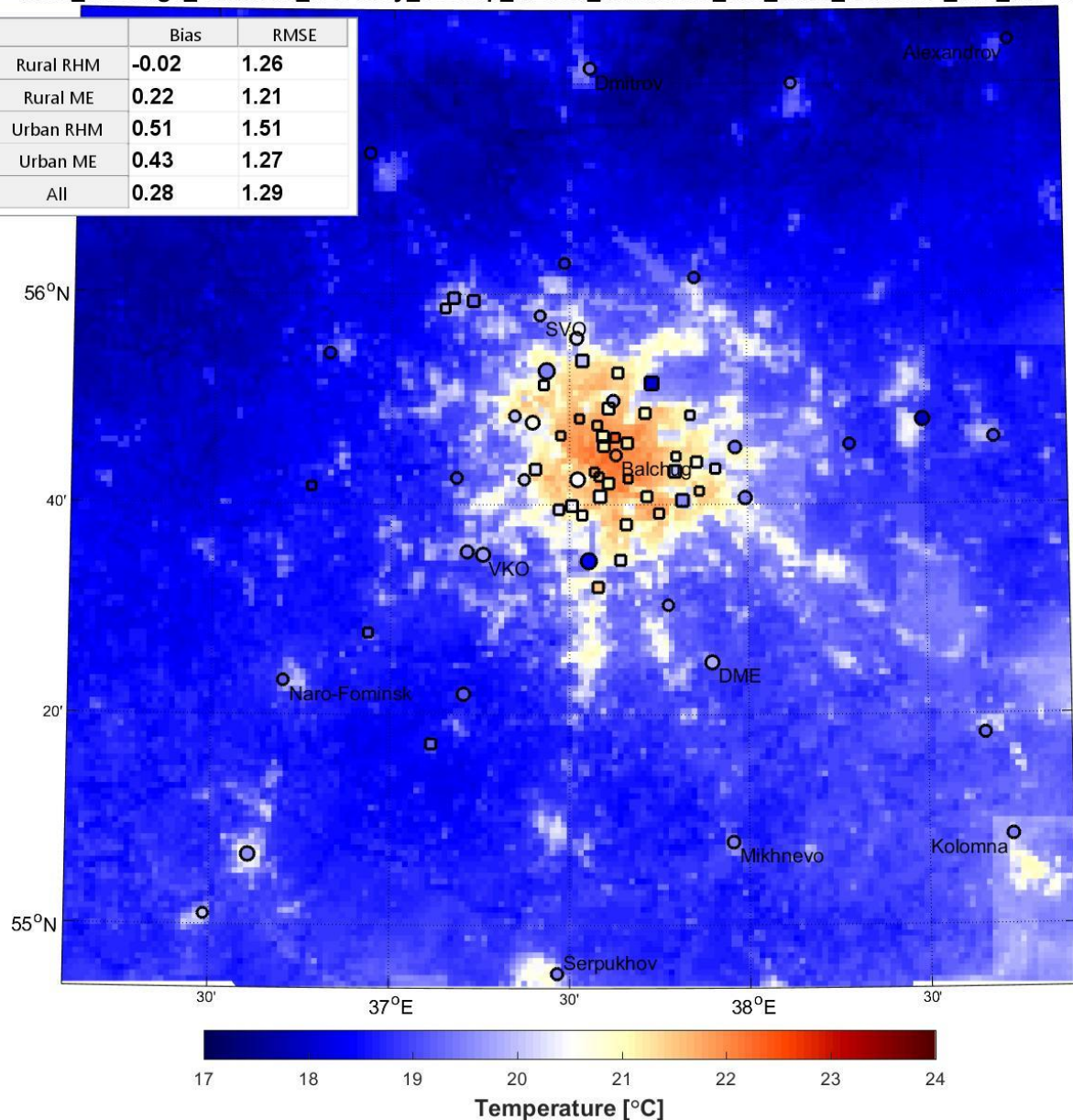
AEV1



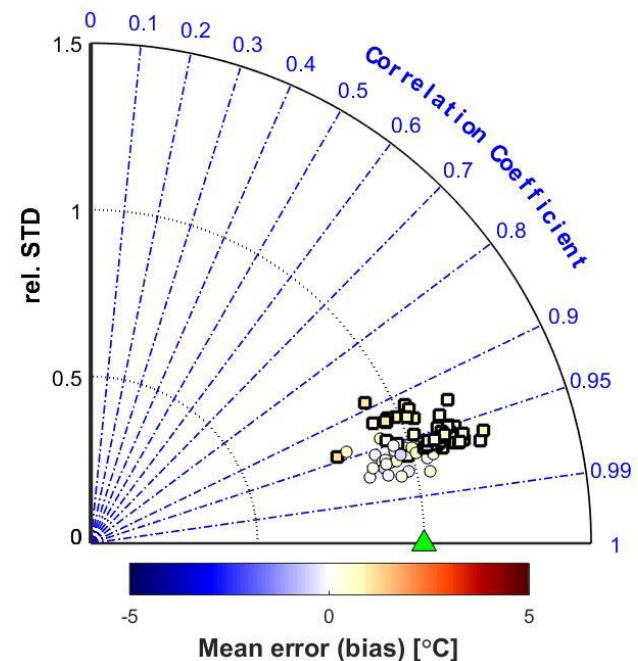
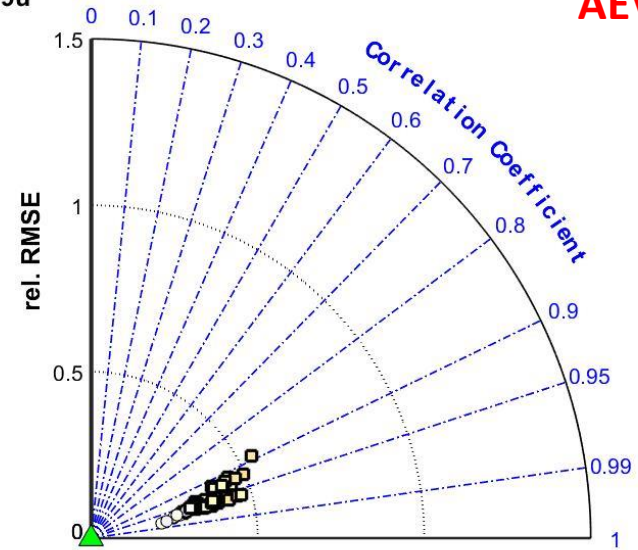
Verification for urban areas: example for case 1

MSK_0.009bg3_osmurb2_v2c Cray_v505up_AEV5a_newturMV_skc_vdif1_osmurb2_v2c_th2 L3n9u

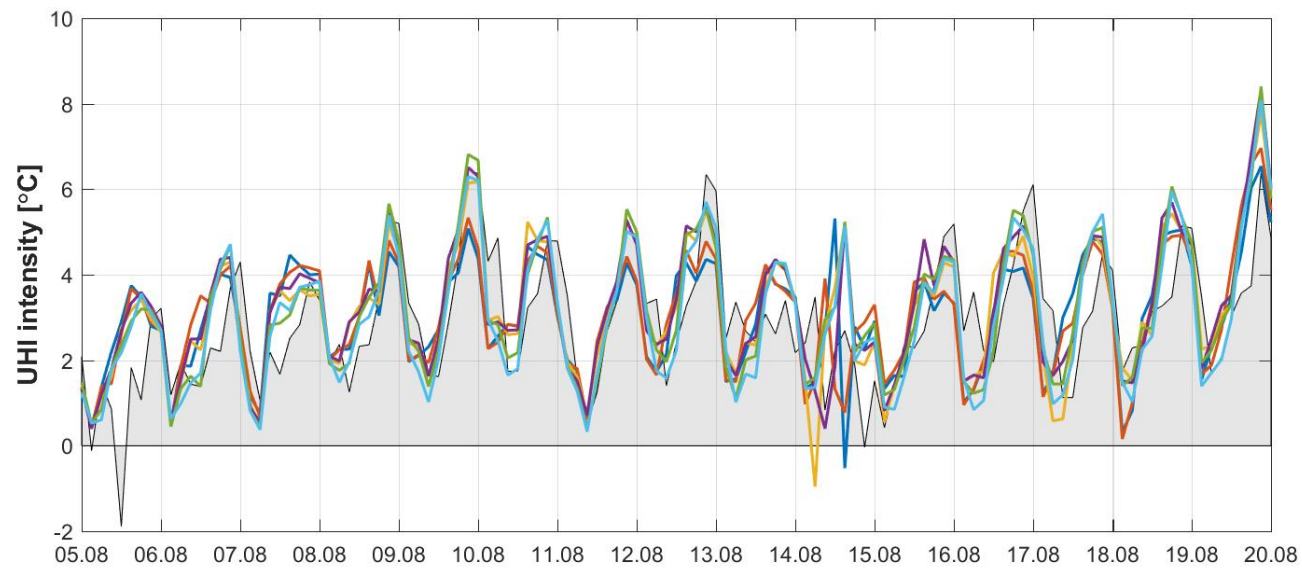
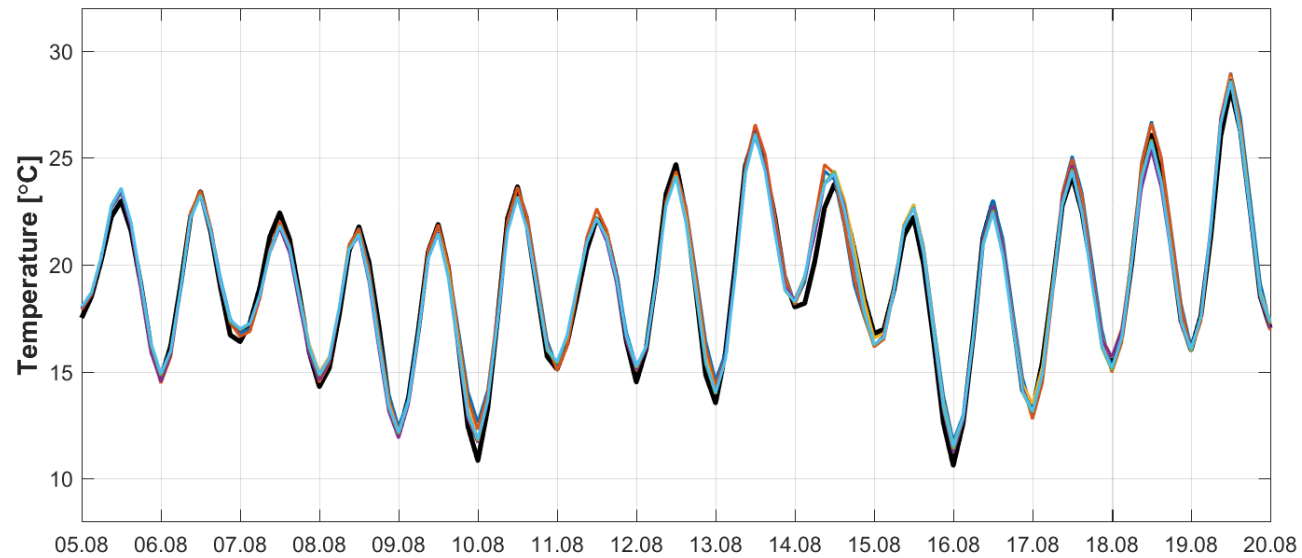
| | Bias | RMSE |
|-----------|-------|------|
| Rural RHM | -0.02 | 1.26 |
| Rural ME | 0.22 | 1.21 |
| Urban RHM | 0.51 | 1.51 |
| Urban ME | 0.43 | 1.27 |
| All | 0.28 | 1.29 |



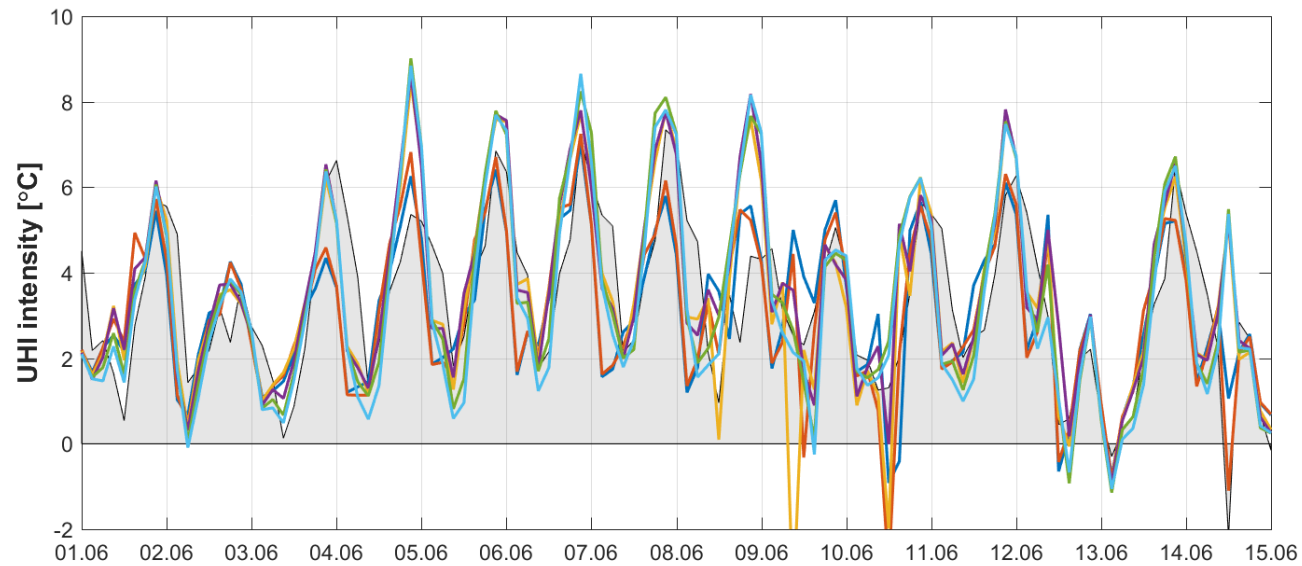
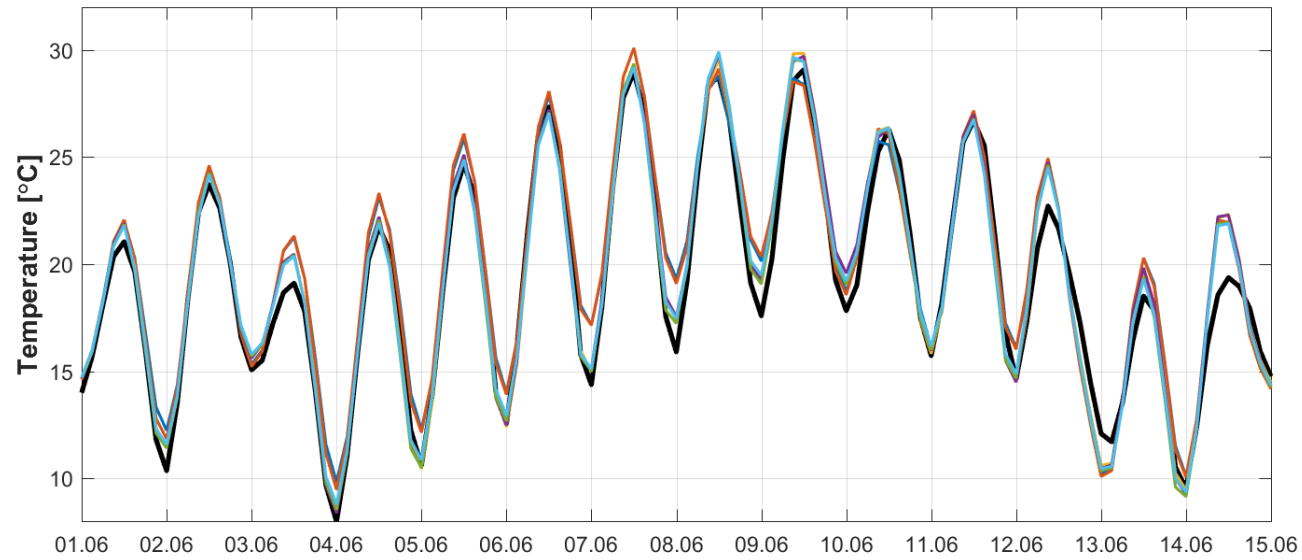
AEV5a



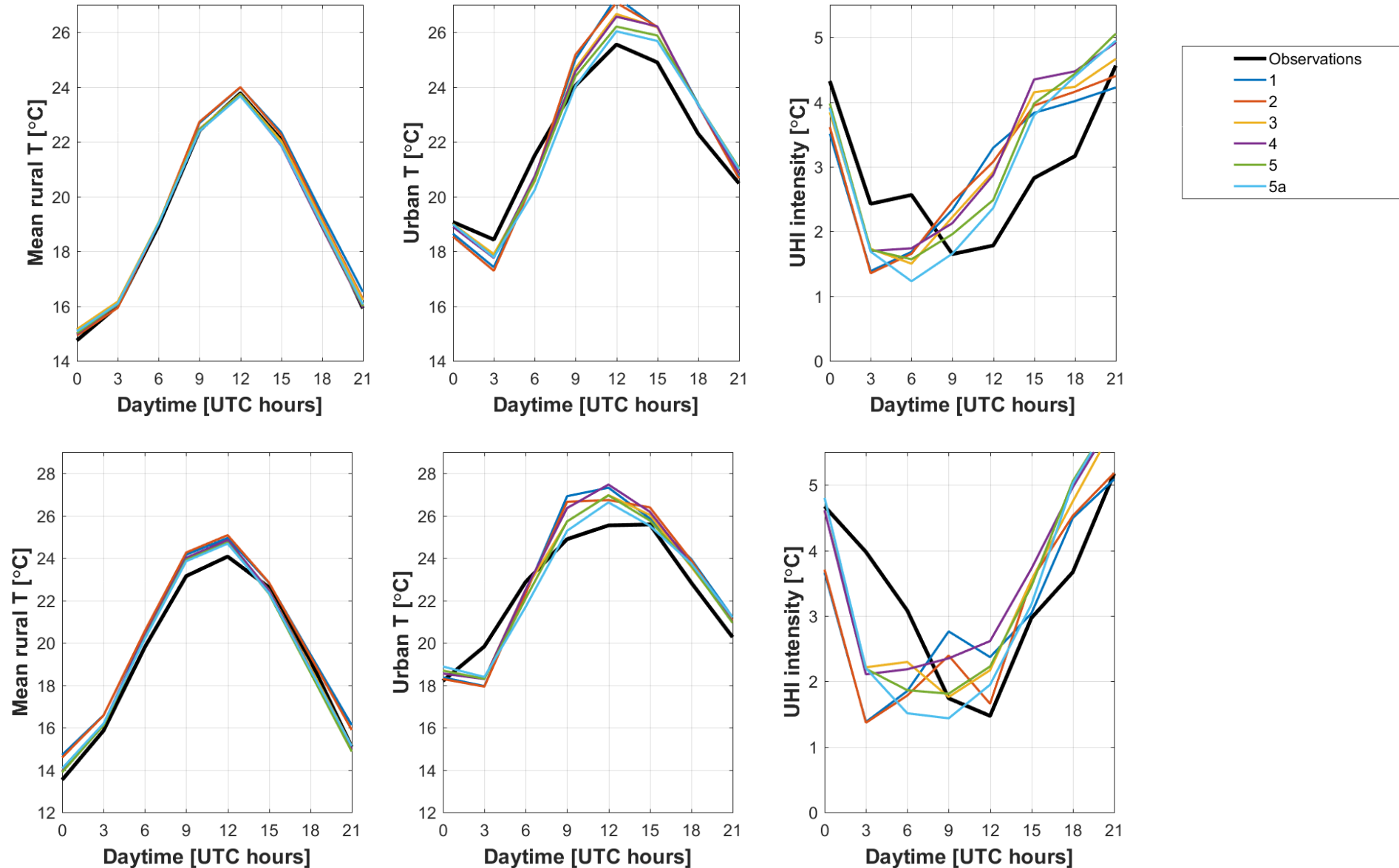
An attempt to compare all runs: case 1



An attempt to compare all runs: case 2



An attempt to compare all runs: case 1 & 2



Verification scores

Case 20170805

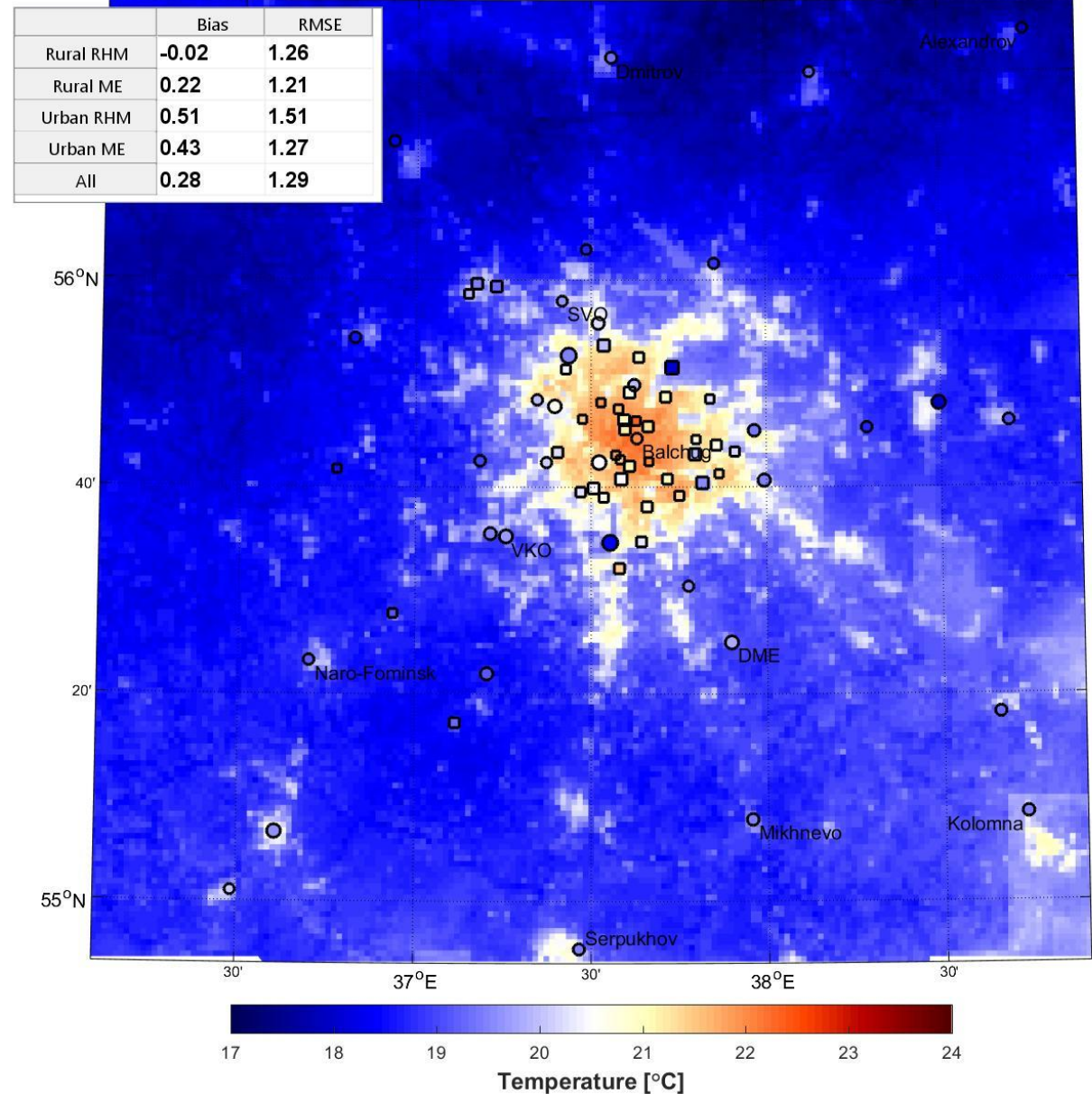
Case 20190601

| | bias (all) | rmse (all) | bias (all) | rmse (all) |
|-------|------------|------------|------------|------------|
| AEV1 | 0.48 | 1.57 | 0.38 | 2.09 |
| AEV2 | 0.45 | 1.6 | 0.36 | 2.09 |
| AEV3 | 0.4 | 1.36 | 0.19 | 1.96 |
| AEV4 | 0.35 | 1.35 | 0.2 | 1.97 |
| AEV5 | 0.3 | 1.31 | 0.07 | 1.95 |
| AEV5a | 0.28 | 1.29 | 0.02 | 1.96 |
| AEV5b | 0.06 | 1.22 | -0.17 | 1.95 |

Surprisingly good results for AEV5b with LCZ-based urban canopy parameters!

AEV5a, case 1

MSK_0.009bg3_osmurb2_v2c Cray_v505up_AEV5a_newturMV_skc_vdif1_osmurb2_v2c_th2 L3n9i



Verification scores

Case 20170805

Case 20190601

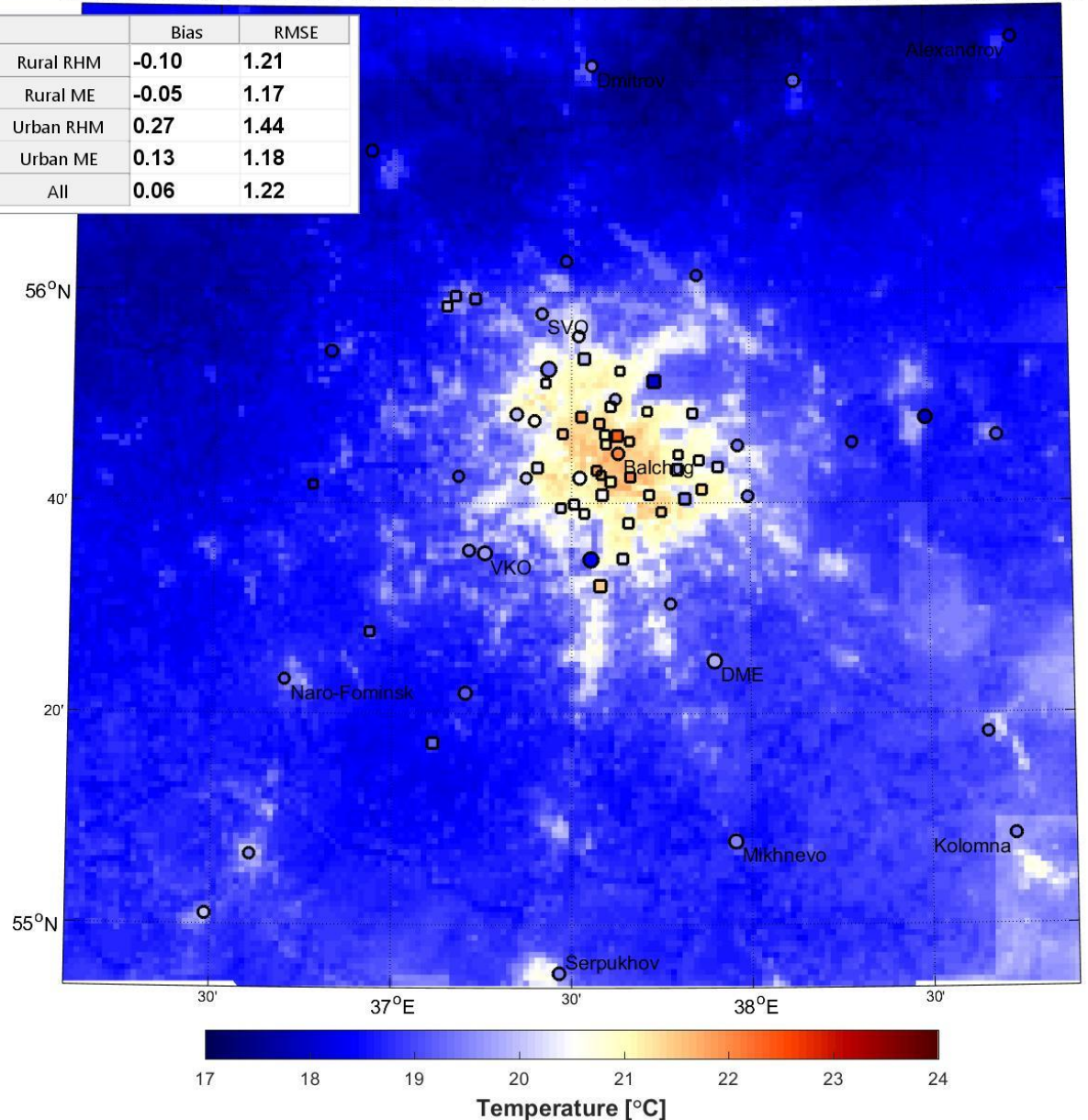
| | bias (all) | rmse (all) | bias (all) | rmse (all) |
|-------|------------|------------|------------|------------|
| AEV1 | 0.48 | 1.57 | 0.38 | 2.09 |
| AEV2 | 0.45 | 1.6 | 0.36 | 2.09 |
| AEV3 | 0.4 | 1.36 | 0.19 | 1.96 |
| AEV4 | 0.35 | 1.35 | 0.2 | 1.97 |
| AEV5 | 0.3 | 1.31 | 0.07 | 1.95 |
| AEV5a | 0.28 | 1.29 | 0.02 | 1.96 |
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Surprisingly good results for AEV5b with LCZ-based urban canopy parameters!

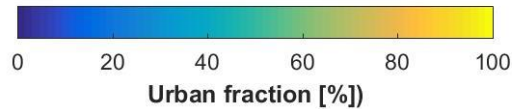
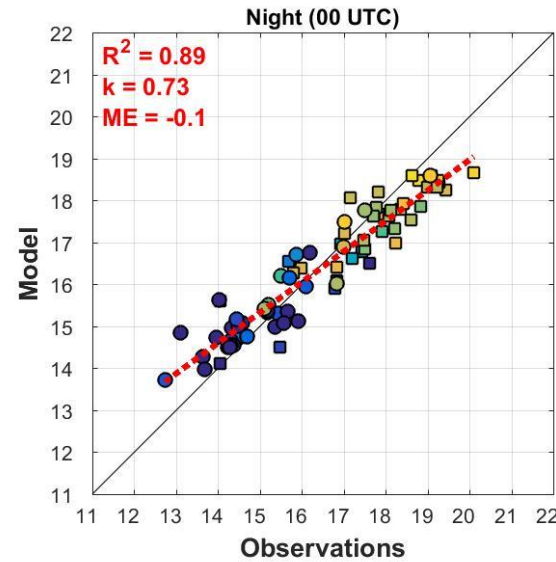
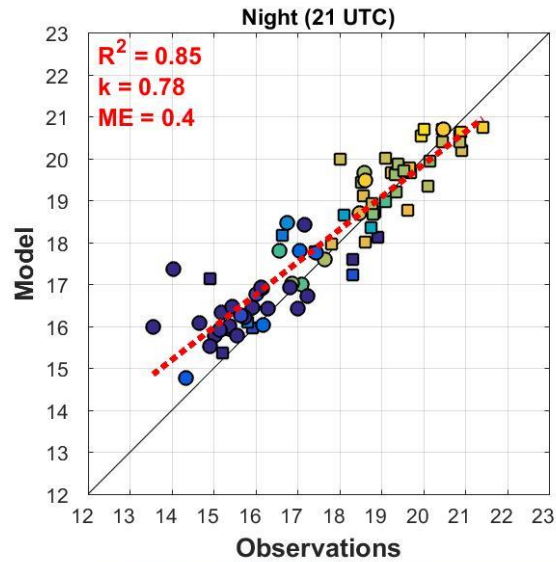
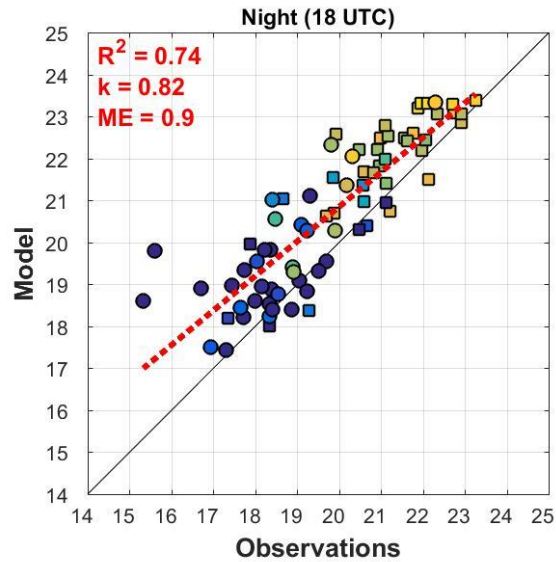
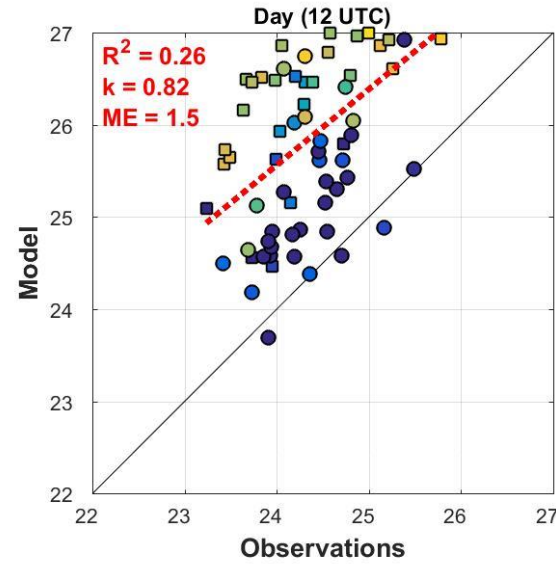
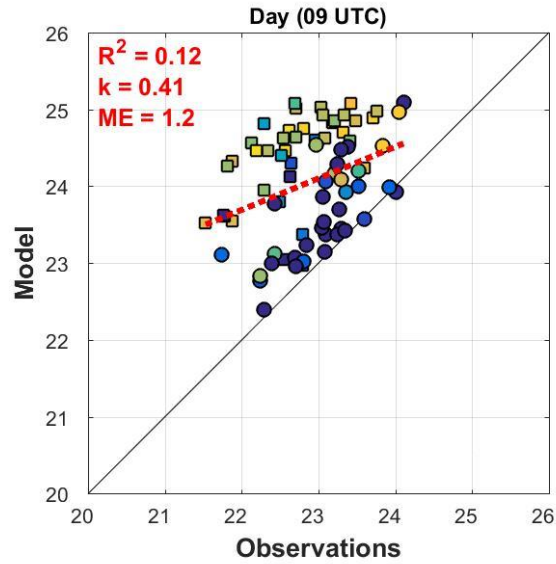
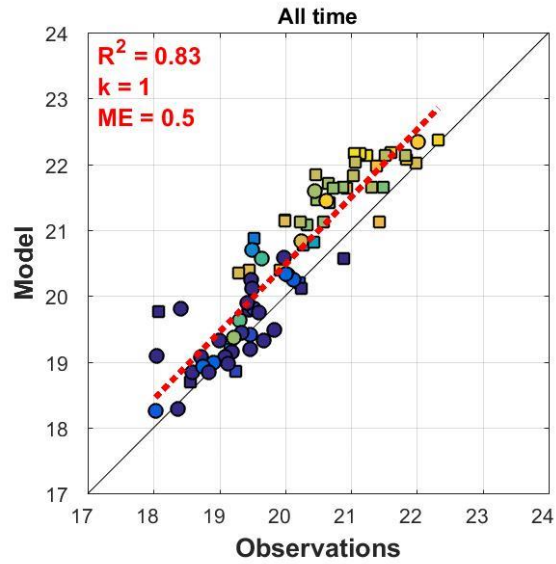
AEV5b, case 1

MSK_0.009bg3_osmurb2_v2c Cray_v505up_AEV5b_newturMV_skc_vdif1_lczdef_th2 L3n9u

| | Bias | RMSE |
|-----------|-------|------|
| Rural RHM | -0.10 | 1.21 |
| Rural ME | -0.05 | 1.17 |
| Urban RHM | 0.27 | 1.44 |
| Urban ME | 0.13 | 1.18 |
| All | 0.06 | 1.22 |



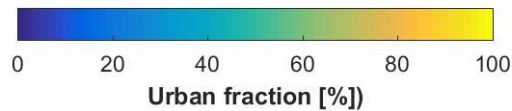
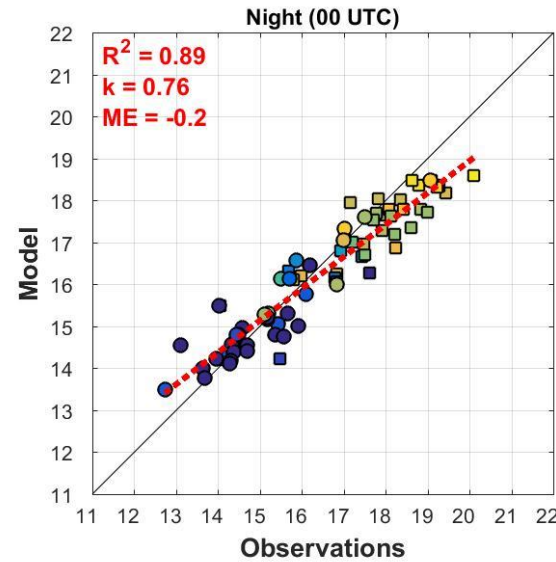
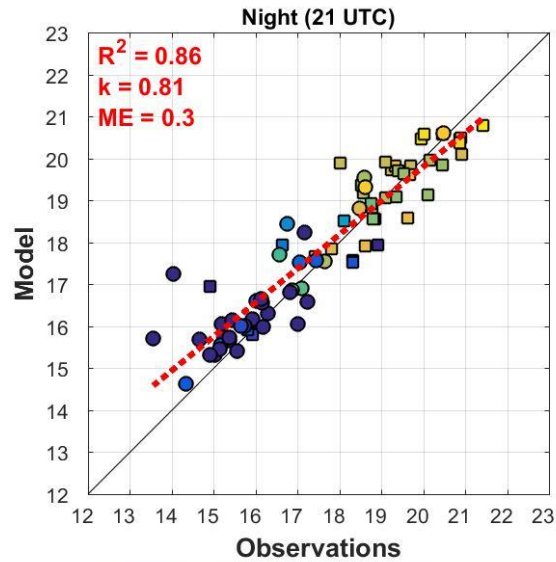
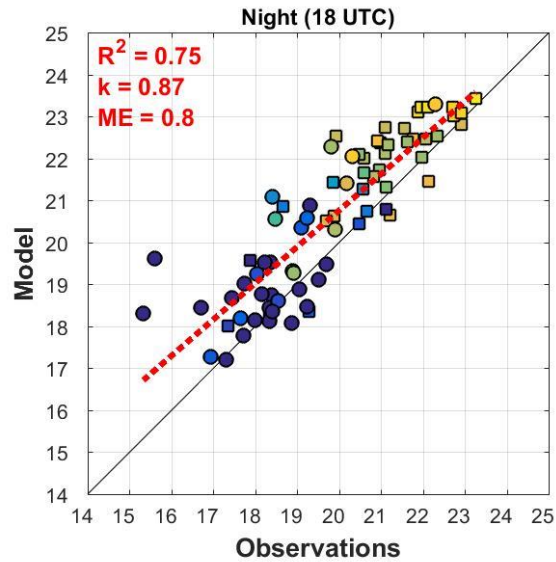
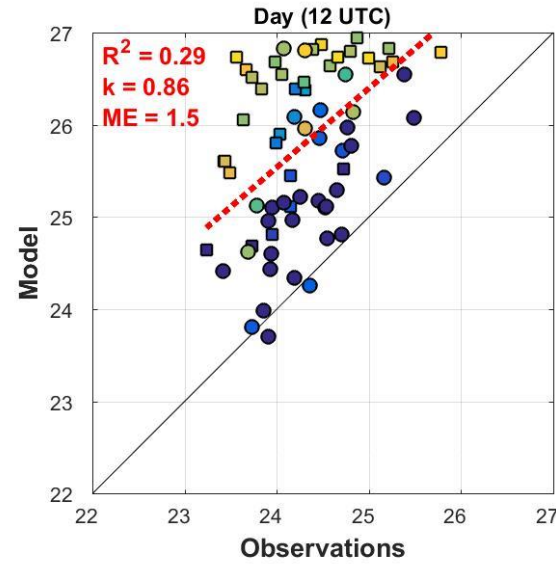
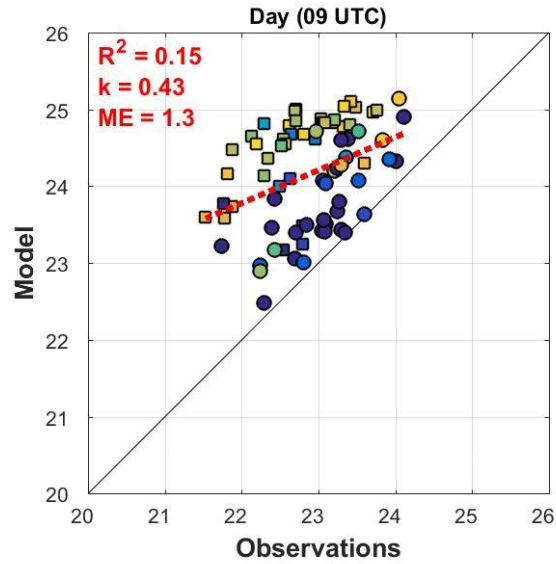
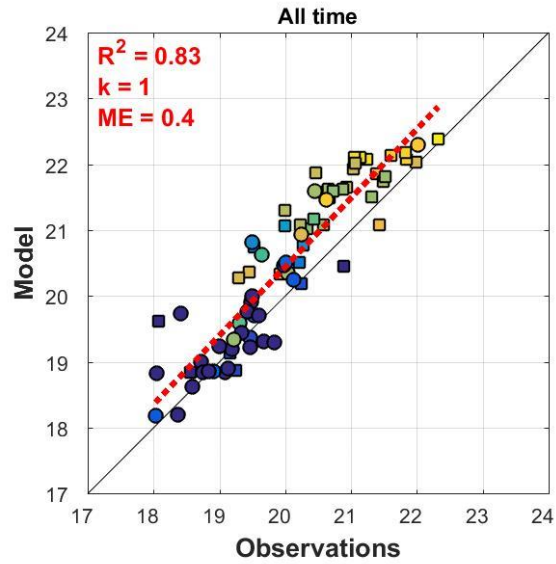
Some more insights



Case 1, AEV1

- `ltype_canopy = 1`
- Old turbulence

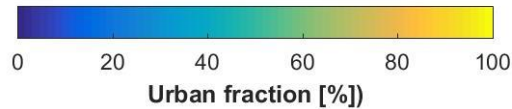
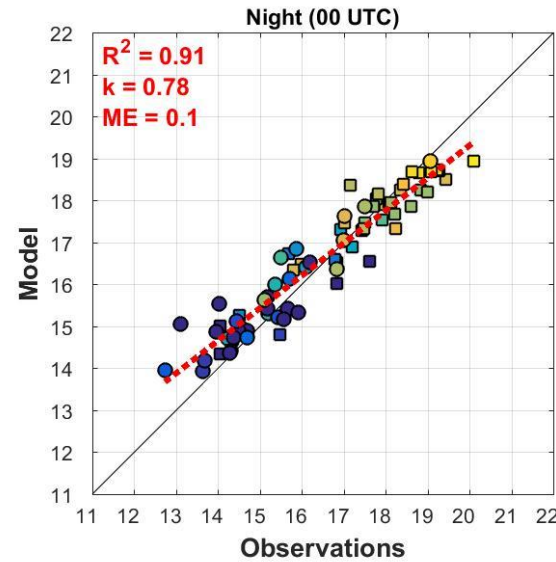
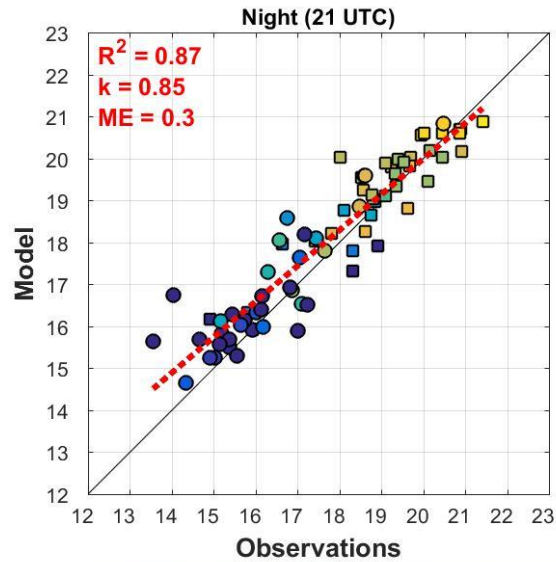
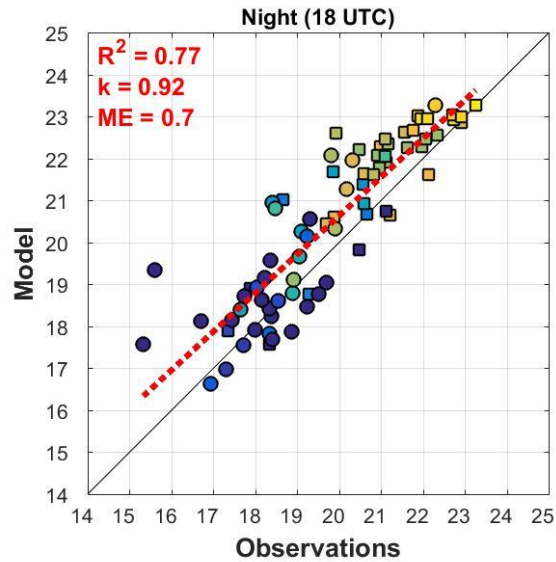
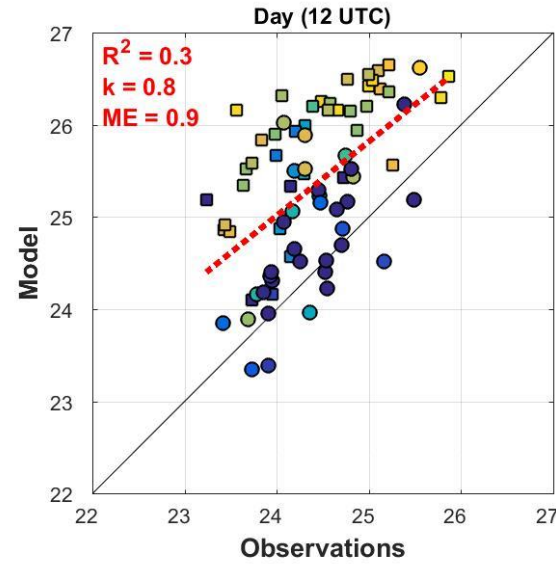
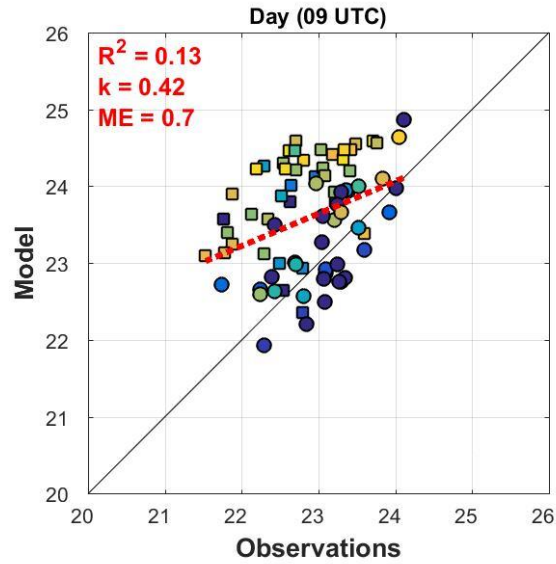
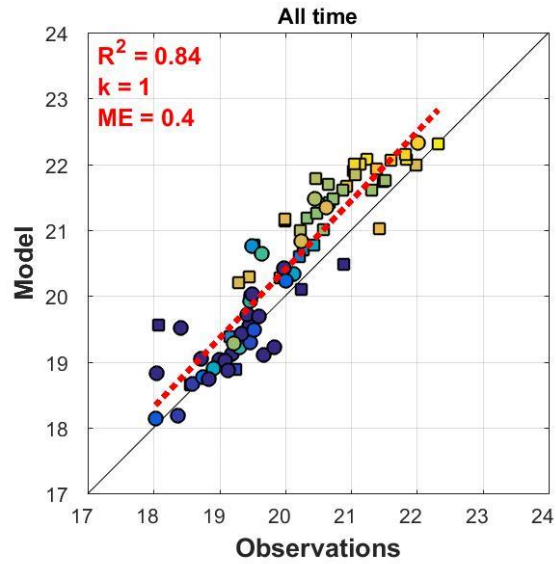
Some more insights



Case 1, AEV2

- Itype_canopy = 2
- Old turbulence

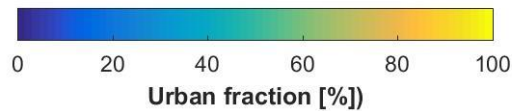
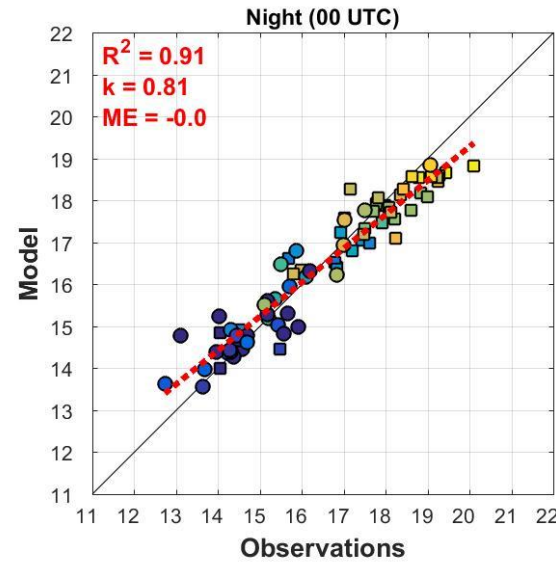
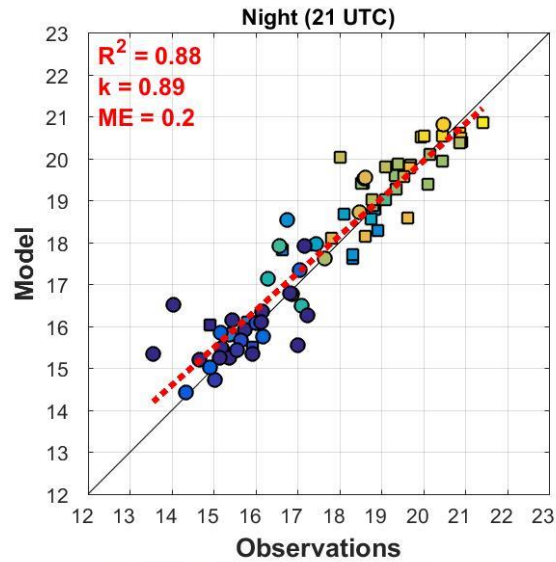
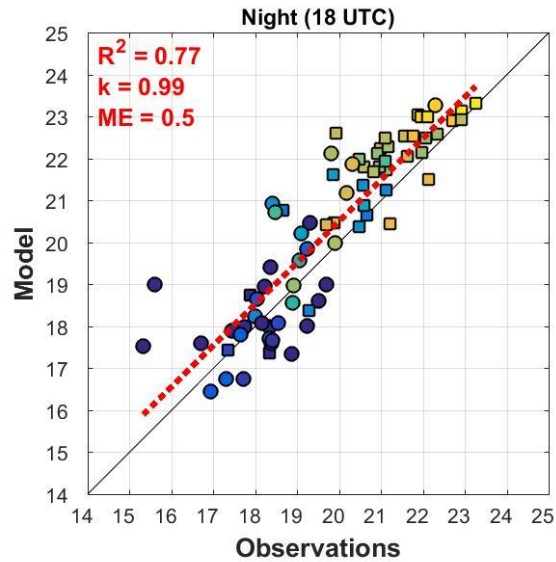
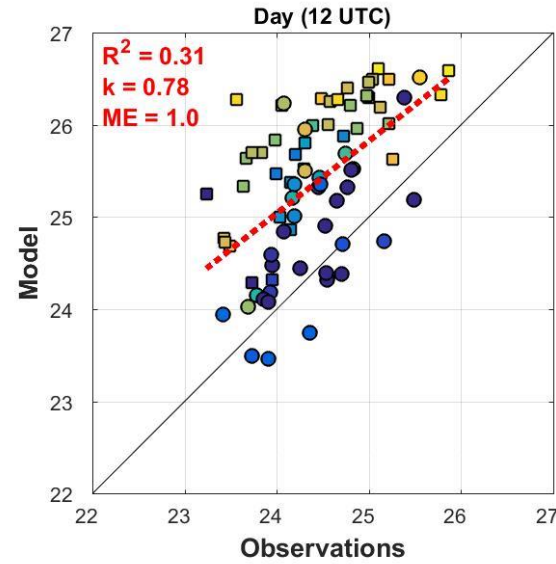
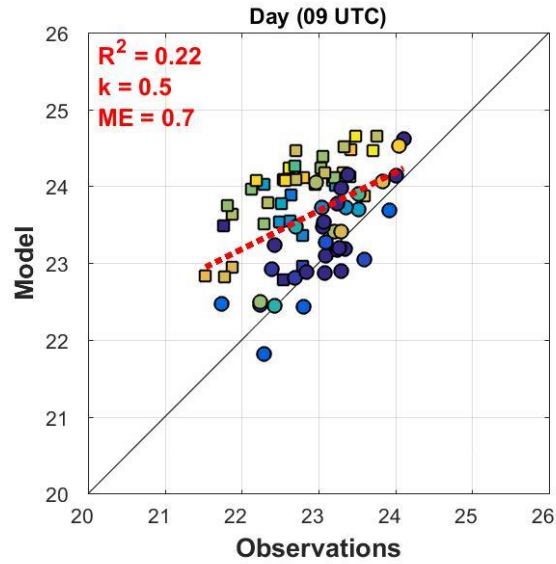
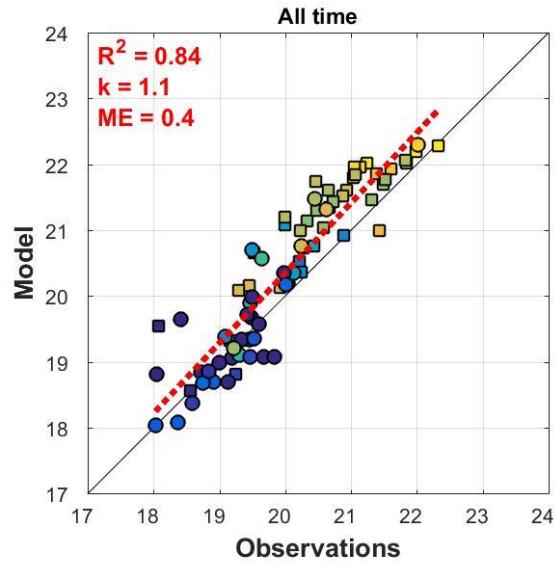
Some more insights



Case 1, AEV3

- `ltype_canopy = 1`
- New turbulence

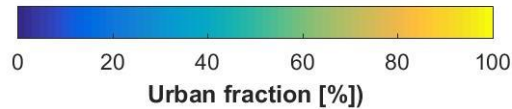
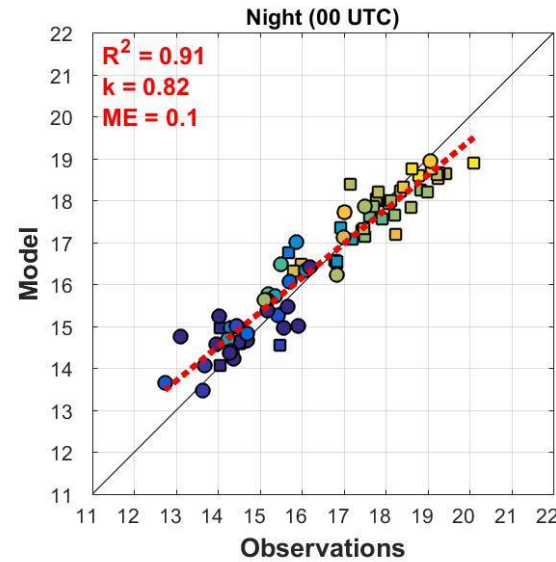
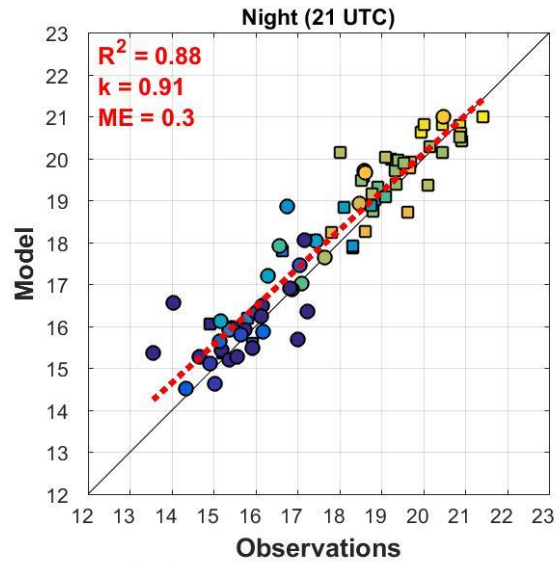
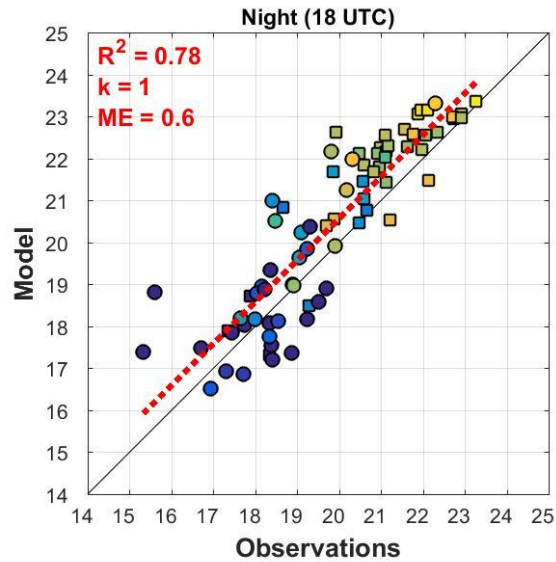
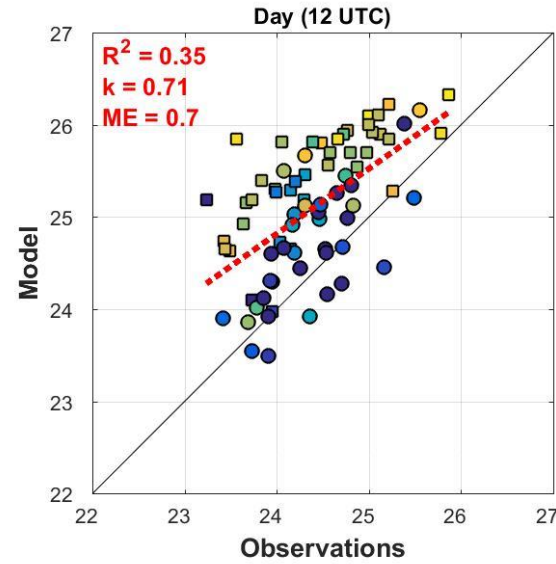
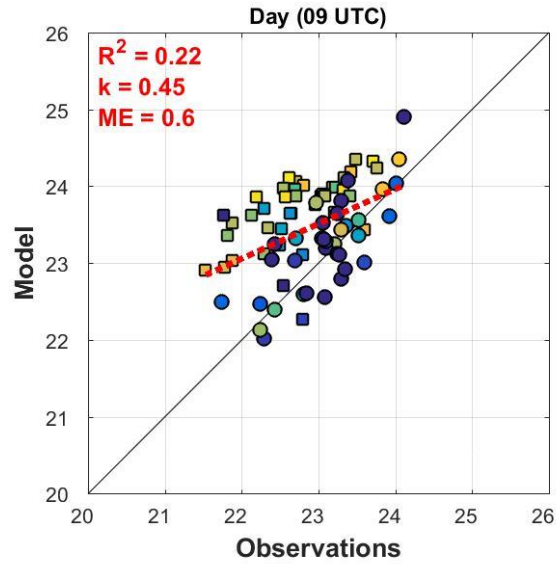
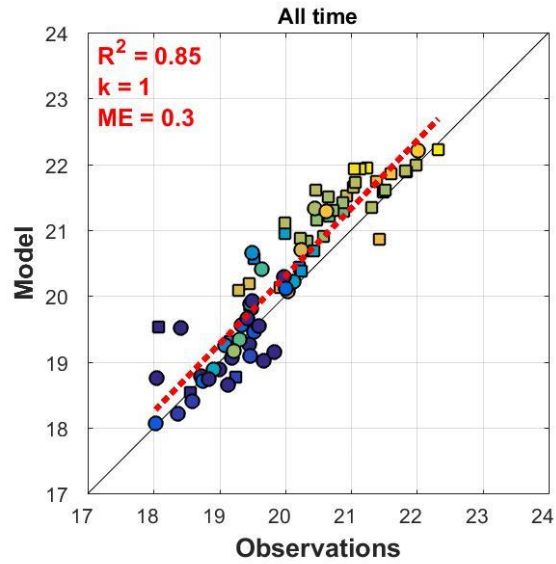
Some more insights



Case 1, AEV4

- Itype_canopy = 2
- New turbulence

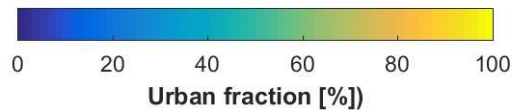
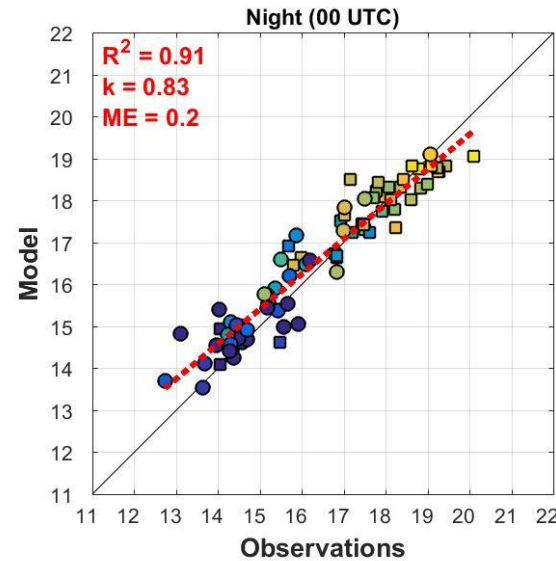
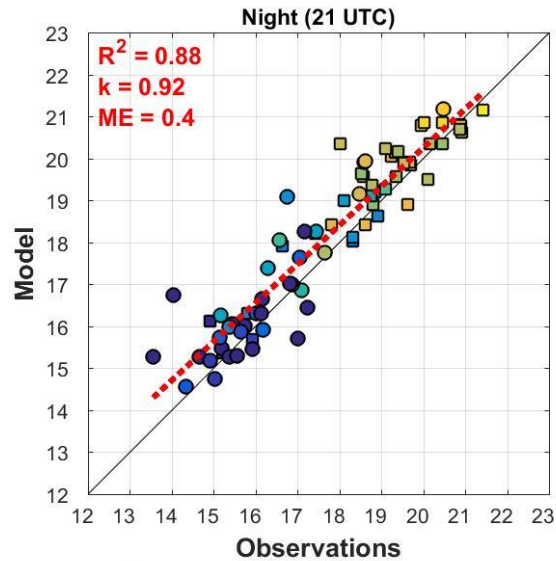
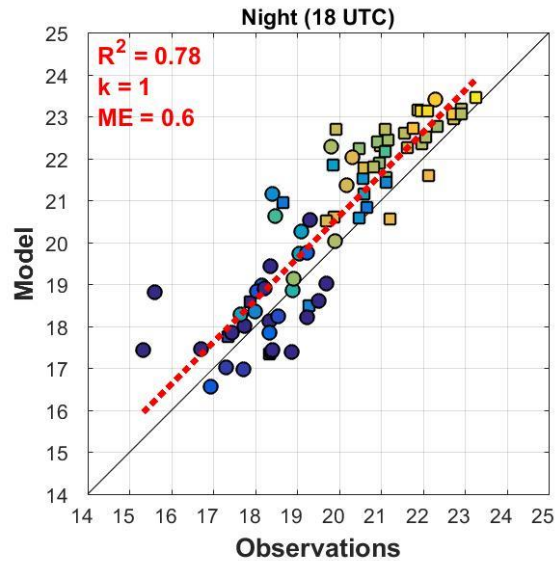
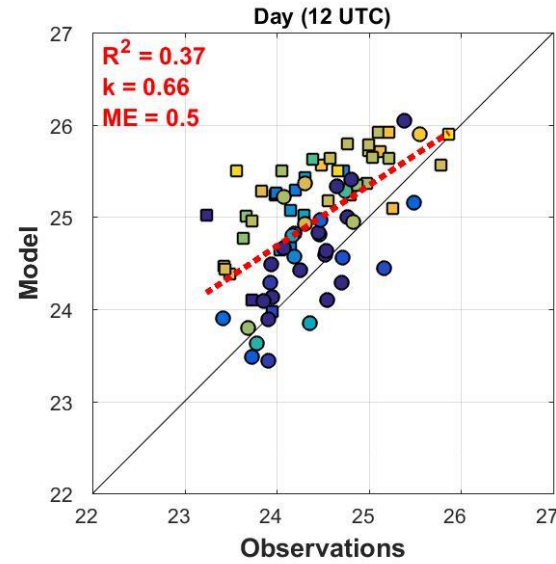
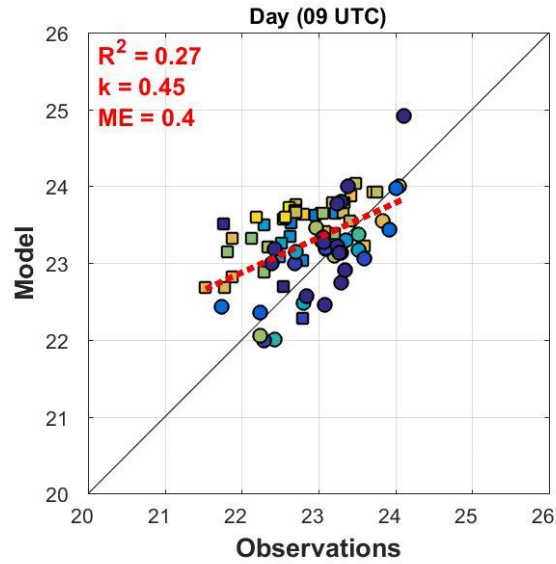
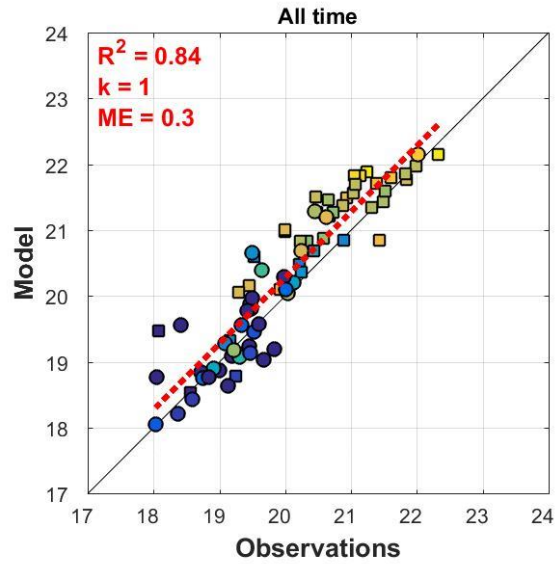
Some more insights



Case 1, AEV5

- $ltype_canopy = 2$
- New turbulence
- $ltype_vdif = 1$,

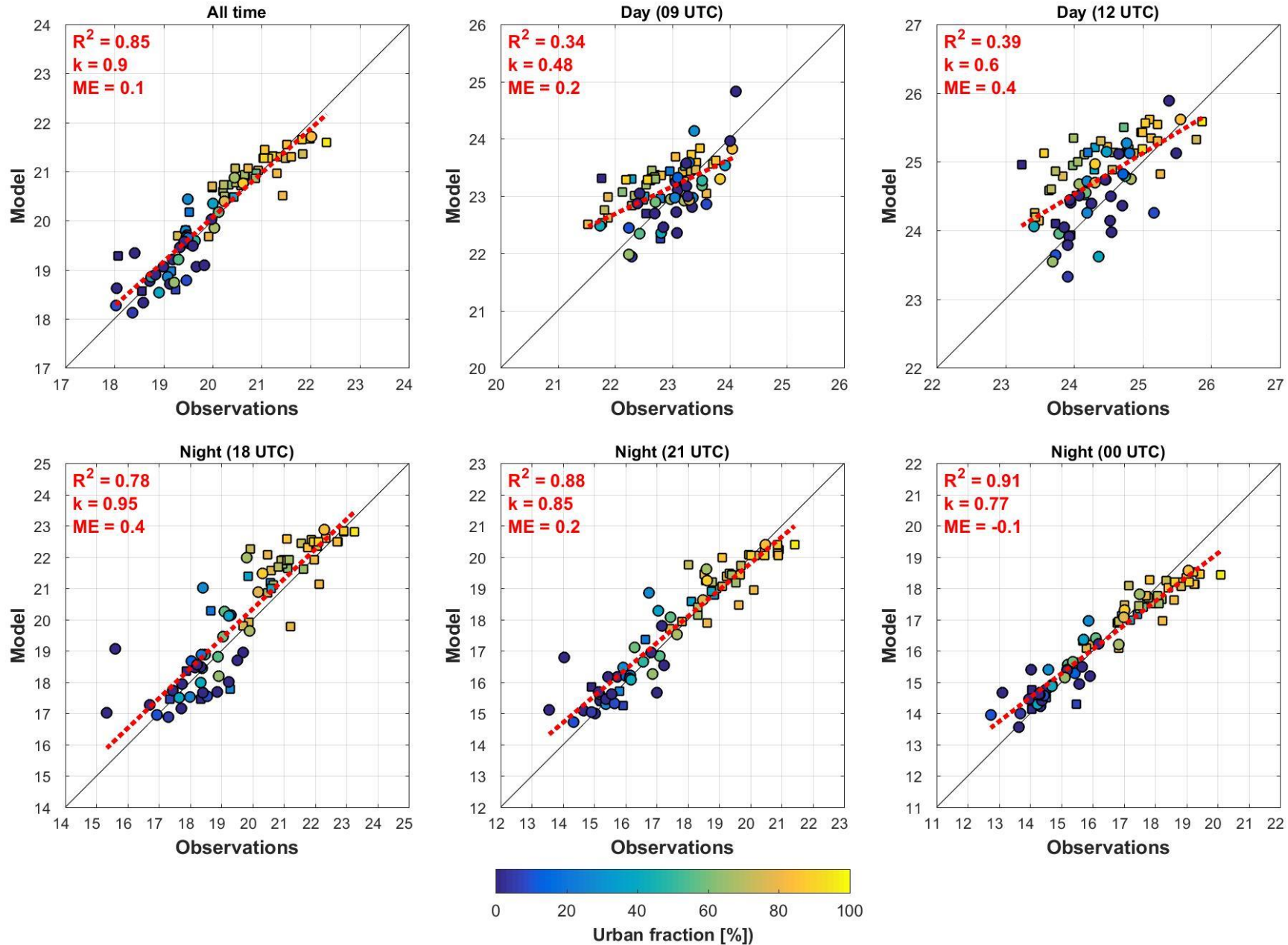
Some more insights



Case 1, AEV5a

- `ltype_canopy = 2`
- New turbulence
- `ltype_vdif=1`,
- Alternative urban thermal parameters

Some more insights



Case 1, AEV5b

- Itype_canopy = 2
- New turbulence
- Itype_vdif=1,
- Alternative urban thermal parameters
- LCZ- based urban canopy parameters

Key conclusions and questions for discussion

- Skin-layer temperature scheme and new turbulence improves modelling results both for rural and urban areas
- `Itype_vdif=1` improves reduces the daytime overheating in the city. Should we include such simulations to the paper?
- Urban thermal parameters needs better calibration. Default values should be discussed with Matthias and Hendrik.
- LCZ-based approach works nice for summer conditions. But what about winter?