

High resolution air temperature maps for urban planning and management

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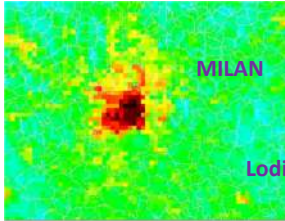
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RATIONALE

- **Urban meteorological networks** are undoubtedly useful but generally **unfit** to climatological studies and unable to describe Air Temperature in the **Urban Canopy Layer** (UCL) with **sufficient spatial resolution**, as required by several professional activities and for local adaptation measures to climate change.
- **Remote sensing data from space** are offering a higher spatial resolution (even if a still very low frequency) of surface characteristics as the **Land Surface Temperature** (LST) and are easily accessible.
- Often used to describe the Surface Urban Heat Islands (S-UHI), **LST** has not a simple correlation with **UCL Air Temperature**, which is the most required variable for planning and management purposes in cities.
- **Using both** high quality in situ measurements of Air Temperature at top of UCL obtained by a dedicated urban network (FOMD CN) and other selected AWS as a **primary variable**,
and satellite derived LSTs as the **secondary one**,
a **Co-Kriging based methodology** has been developed and tested
to obtain medium to **high spatial resolution UCL Air Temperature maps**
- **Instantaneous** as well as **mean fields of fine spatially resolved Air Temperature** find relevant application not only in monitoring and assessing activities of adaptation and mitigation measures in the urban environment, but also in urban climate studies.

Workflow

- **1st Variable:** T_{air} measured by surface meteorological networks in the UCL (z_1)



- **2nd Variable:** LST obtained by IR radiation remotely sensed from space (z_2)

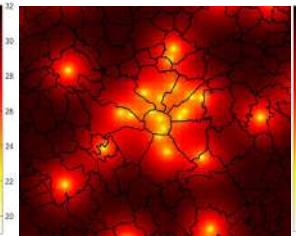
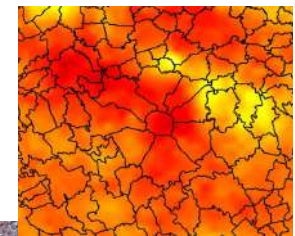
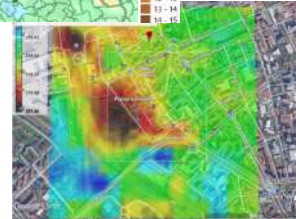
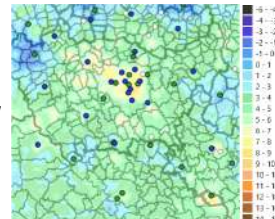
Co-Kriging of 1st (primary) and 2nd (secondary) variables:

$$z_1^*(x_0) = \sum_{\alpha=1}^{n_1} \lambda_{\alpha} \cdot z_1(x_{\alpha}) + \sum_{\alpha=1}^{n_2} \omega_{\alpha} \cdot z_2(x_{\alpha})$$

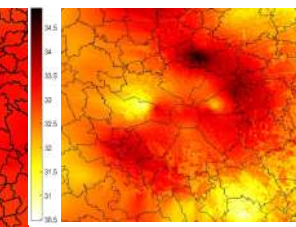
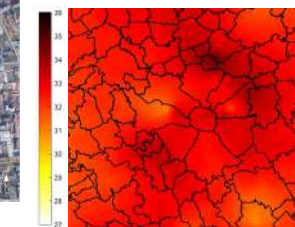
⇒ **Results:** Medium- to high-resolution T_{air} fields in UCL and associated uncertainties

⇒ **Applications:**

- Climatology of UHI and HW
- Air Temperature Atlas
- Assessment of urbanistic modifications for adaptation to Climate Change



Medium resolution
($10^2\text{m} \times 10^2\text{m}$) COK
interpolation:
 T_a and $U(T_a)$
2018-07-07: 22UTC



Medium (left)
and high (right:
 $30\text{ m} \times 30\text{ m}$) COK
interpolated T_a :
2019-07-26:10utc

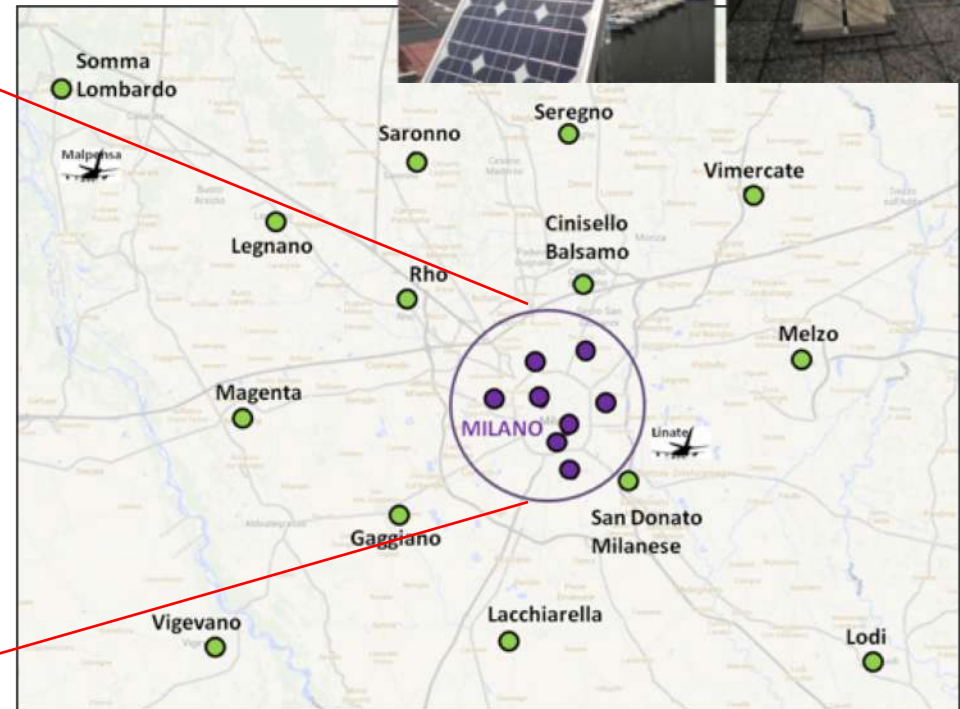
Primary Variable (1):

T_{air} from a high quality meteorological network (under sampled)

1) FOMD Italian Urban Climate Network (CN), Milan subnet (21 AWS)

(Curci et al., MST, 2017):

T_{air} and other ECVs at top of UCL in Milan and surroundings measured with metrological criteria since 2011 (*MeteoMet Project*)



Primary Variable (2):

T_{air} from supplementary meteorological networks (under sampled)

- Selected (Frustaci et al., EGU 2019) stations for T_{air} :

- in the city (urban Air Quality network)
- and outside the city (rural stations)

from **Regional Meteorological Service**

(26 AWS, ARPA Lombardia:

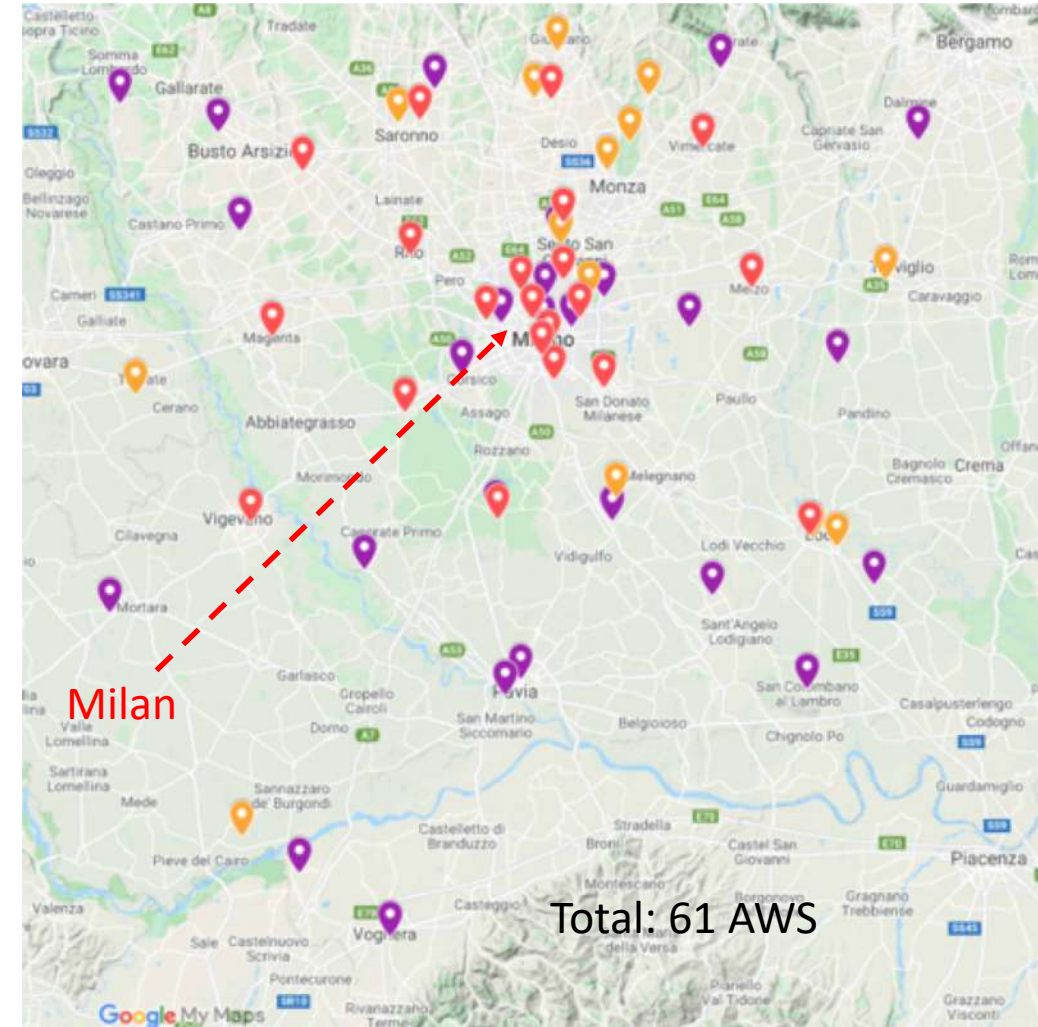
<https://www.arpalombardia.it/>):

- Selected stations for T_{air} from amateur network, 14 AWS)

(Meteonetwork: <https://www.meteonetwork.it/>):

Climate Network – FOMD (CN, 21 AWS):

(<https://www.fondazioneomd.it/>)



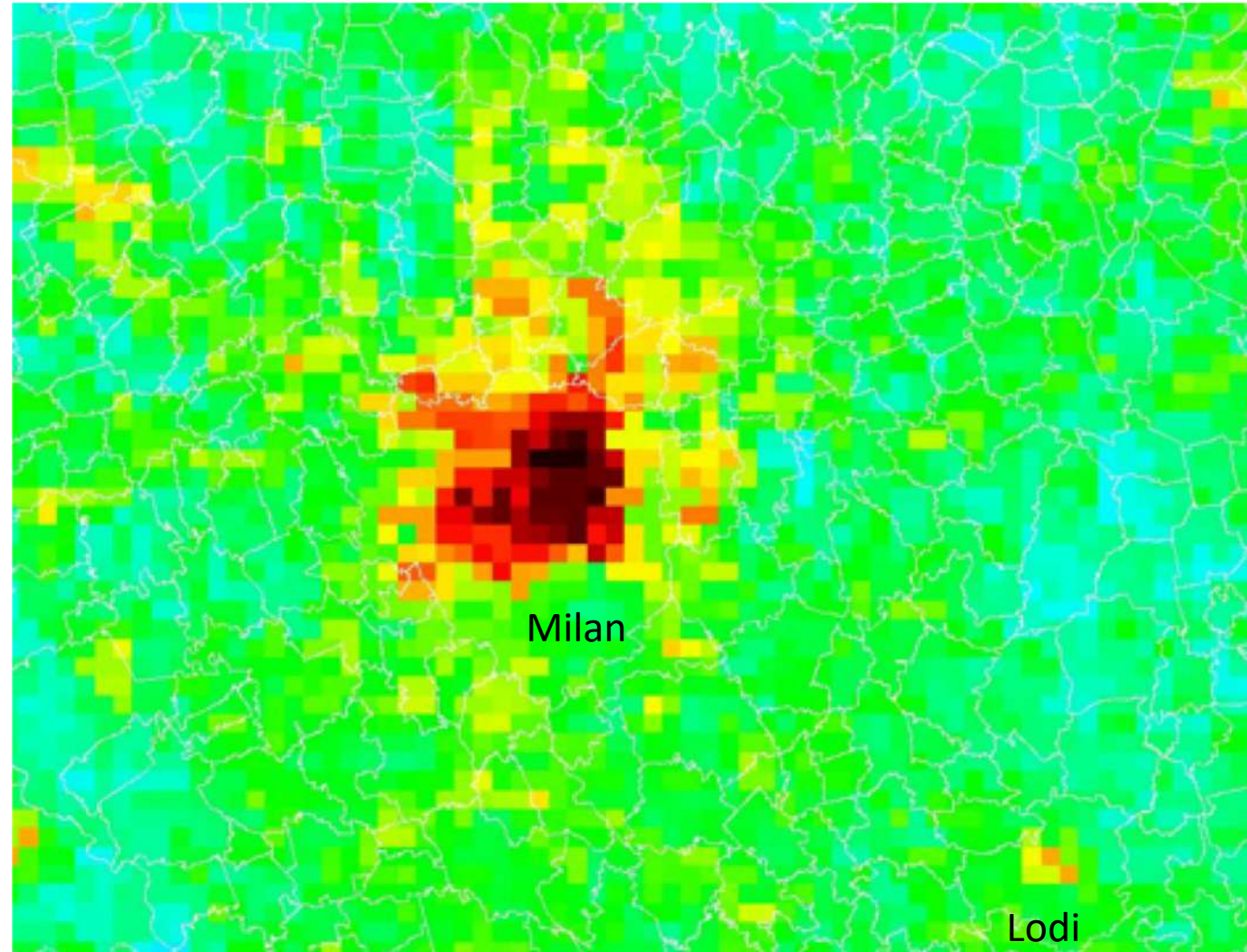
Secondary Variable (1): T_s remotely sensed from space (LST)

- Land Surface Temperature (LST)
at medium resolution (1000 m)
from Sentinel 3 A&B— ESA-Copernicus

(Sobrino et. al., Remote Sensing of Env., 2016)

Good spatial sampling!

Sentinel 3 (LST, 1 km resolution):
Milan and surroundings, 2018-07-07 09:10 UTC

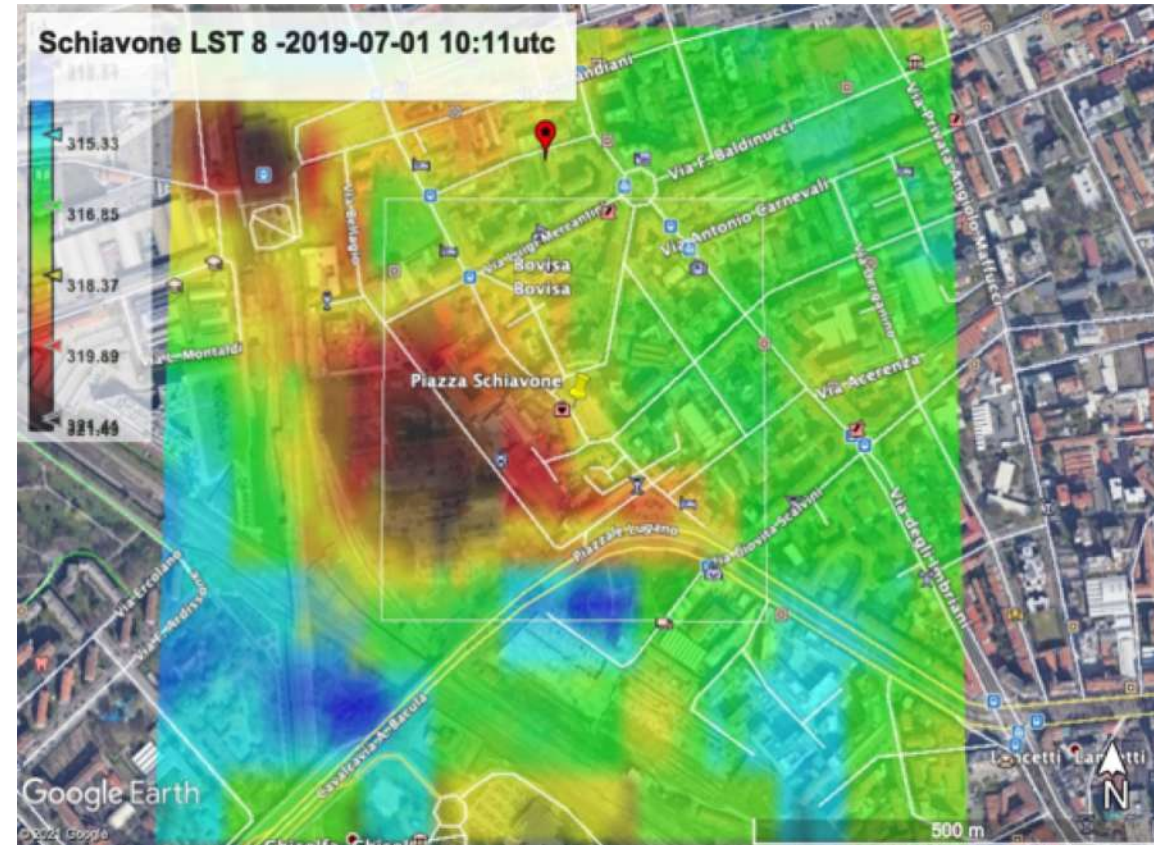


Secondary Variable (2): T_s remotely sensed from space (LST)

- Land Surface Temperature (LST)
at high resolution (30 m)
from Landsat 8 (NOAA & USGS)

(Ermida et al., Remote Sensing 2020)

Very good spatial sampling!



Landsat 8 (LST, 30 m resolution):
Milan, Piazza Schiavone and surroundings, 2019-07-01 10:11 UTC

STEP 1

Empirical **semi-variograms** γ for T_{air} and for **LST** data:

$$\gamma(h) = \frac{1}{2N(h)} \sum_{i=1}^{N(h)} [z(x_i) - z(x_i + h)]^2$$

- **Z**: regionalized variable
- x_i : position in the domain
- **h**: distance between points
- **N(h)**: number of point pairs separated by **h**.

Kriging of T_{air} and **LST**: co-kriging (1)

STEP 2

Fitting of a model for the semi-variogram
(a *spherical model* proves to be the best one):

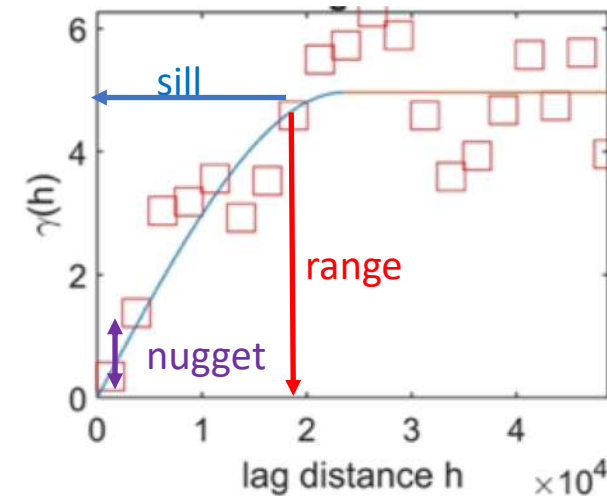
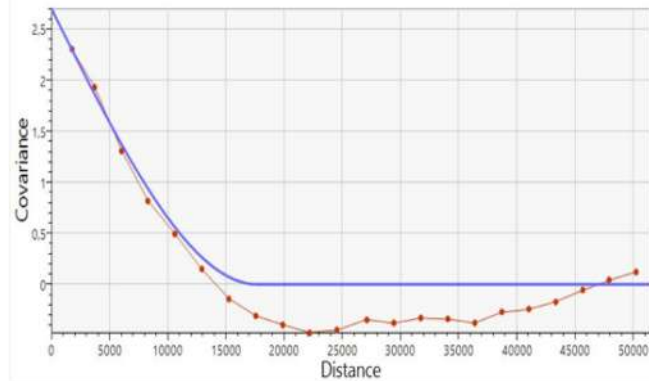
$$\gamma(h) = \begin{cases} c \left[1,5 \cdot \left(\frac{r}{a} \right) - 0,5 \cdot \left(\frac{r^3}{a^3} \right) \right] & r = |h| \leq a \\ c & r = |h| \geq a \end{cases}$$

STEP 3

Cross-variogram
(build with the covariances of T_{air} and **LST** and fitted again with a spherical model)

$$\gamma'(h) = \frac{1}{N(h)} \sum_{i=1}^{N(h)} [z_1(x_i) - Z_1][z_2(x_i + h) - Z_2]$$

Z_1, Z_2 : Mean values



⇒ **a**: Range

⇒ **c**: Sill

(Nugget ≈ 0)

Kriging of T_{air} and LST: co-kriging (2)

STEP 4: co-kriging Interpolation

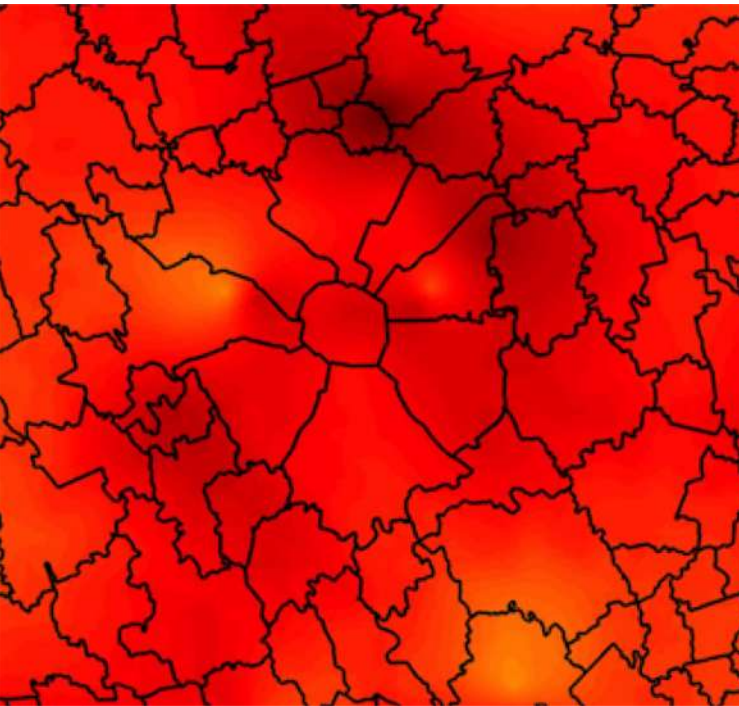
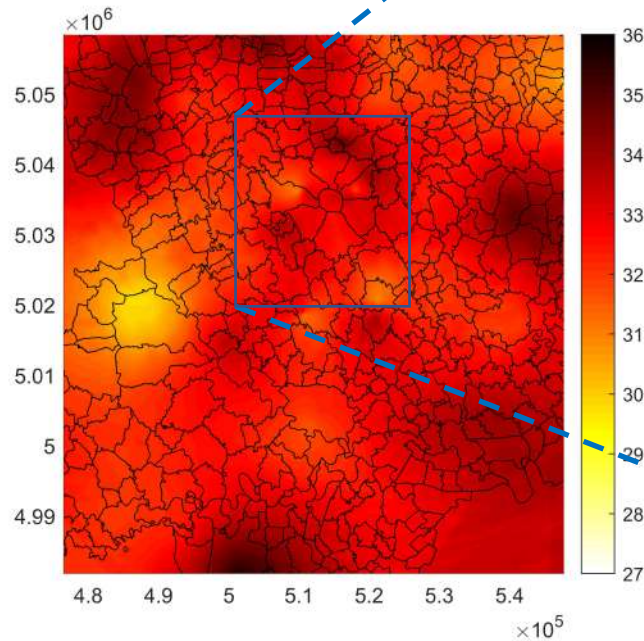
$$z_1^*(x_0) = \sum_{\alpha=1}^{n_1} \lambda_{\alpha} \cdot z_1(x_{\alpha}) + \sum_{\alpha=1}^{n_2} \omega_{\alpha} \cdot z_2(x_{\alpha})$$

where z_1 is T_{air} and z_2 is the LST value (λ_{α} and ω_{α} are the correspondent weights)

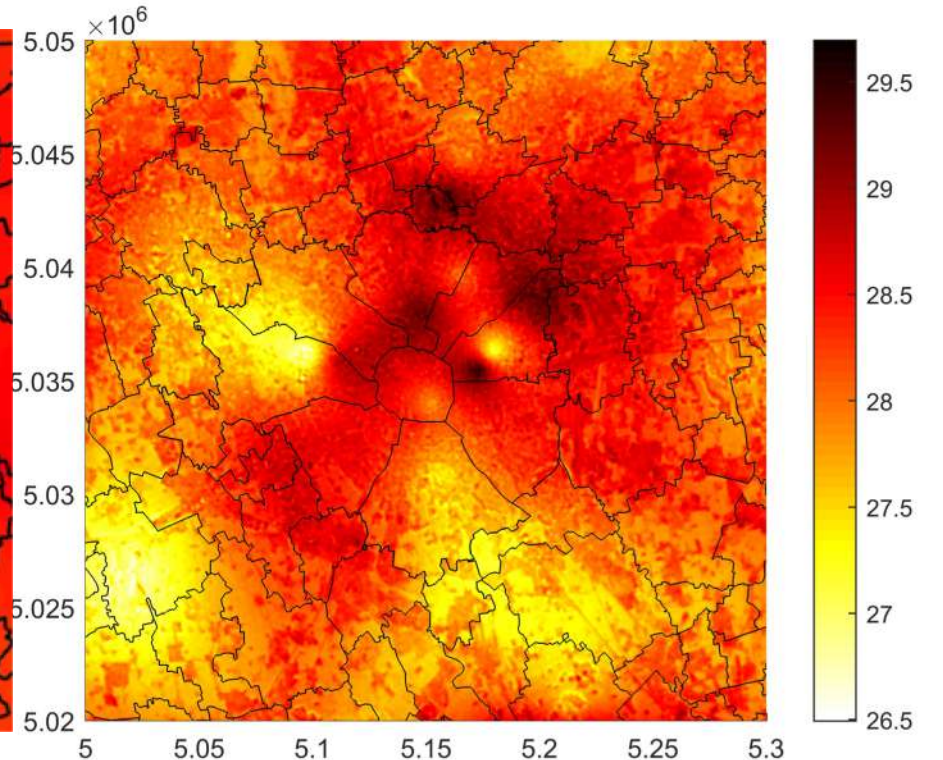
The weights are obtained as solutions of a linear equation system (Linear Model of Coregionalization), with the condition of positive definiteness (Bourgault et al. 1991)

Example: Heat Wave episode and Milan UHI at 100 m and at 30 m resolution

2019-07-26, AM:
Heatwave episode with
the Milan UHI maximum
shifted to NE



Cokriging T_{air} field at **100 m**
using LST data from Sentinel 3



Cokriging T_{air} field at **30 m**
using LST data from Landsat 8

Selection Criteria:

- $I_{UHI} < 3^{\circ}\text{C}$: Well developed Urban Heat Island over the city
(at least one of eight downtown AWS)
- **Max (I_{uhi}) position in the city**: the 6 most frequent choices (**NE, E, S, W, NW, Centre**)
- $V_x < 1.3 \text{ m/s}$: low winds, relative to the local climatology (Po valley)
(Frustaci et al., Climatology of the Milano Canopy Urban Heat Island by means of an operational urban meteorological network, 2° National Congress AISAM, Naples, 2019)
- **Weather Type**: high pressure anticyclonic (Borghi S., Giuliacci M.,..... , 1979)
- **Heat Waves** (WMO, Heatwaves and Health: Guidance on Warning-System Development, 2015)

Summer mornings:
None (no $I_{uhi} \geq 3^{\circ}\text{C}$)

Further:

- Only **Winter** (DJF)
and **Summer** (JJA) months
- Only **Morning** (10 to 12 LT)
and **evening** (21 to 23 LT)
hours (Satellite passes)

Zone	No of events	Percentage (%)
NE	1	0,5
Centre	166	79,4
NW	0	0,0
E	39	18,7
W	3	1,4
S	0	0,0
TOTAL	209	100

Winter evenings

Zone	No of events	Percentage (%)
NE	16	17,8
Centre	41	45,6
NW	3	3,3
E	14	15,6
W	14	15,6
S	2	2,2
TOTAL	90	100

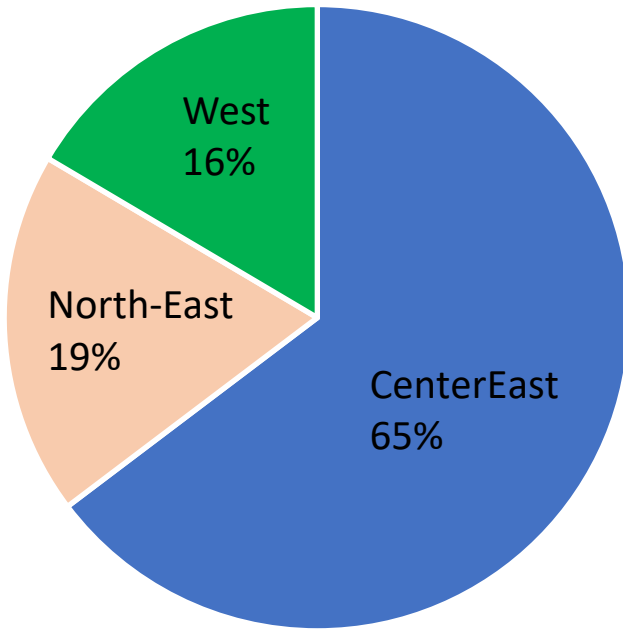
Winter mornings

Zone	No of events	Percentage (%)
NE	8	2,3
Centre	202	58,4
NW	68	19,7
E	61	17,6
W	2	0,6
S	5	1,4
TOTAL	346	100

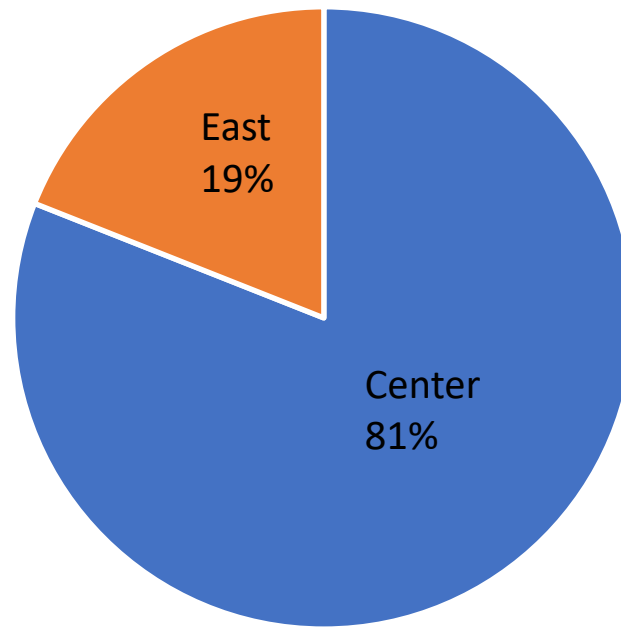
Summer evenings

Frequencies of selected Weather Types for applications

Winter UHI 11:00 UTC+1

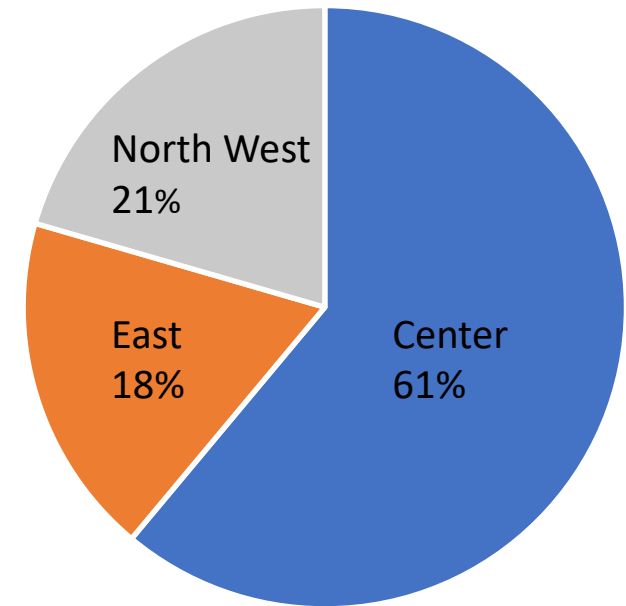


Winter UHI 22:00 UTC+1



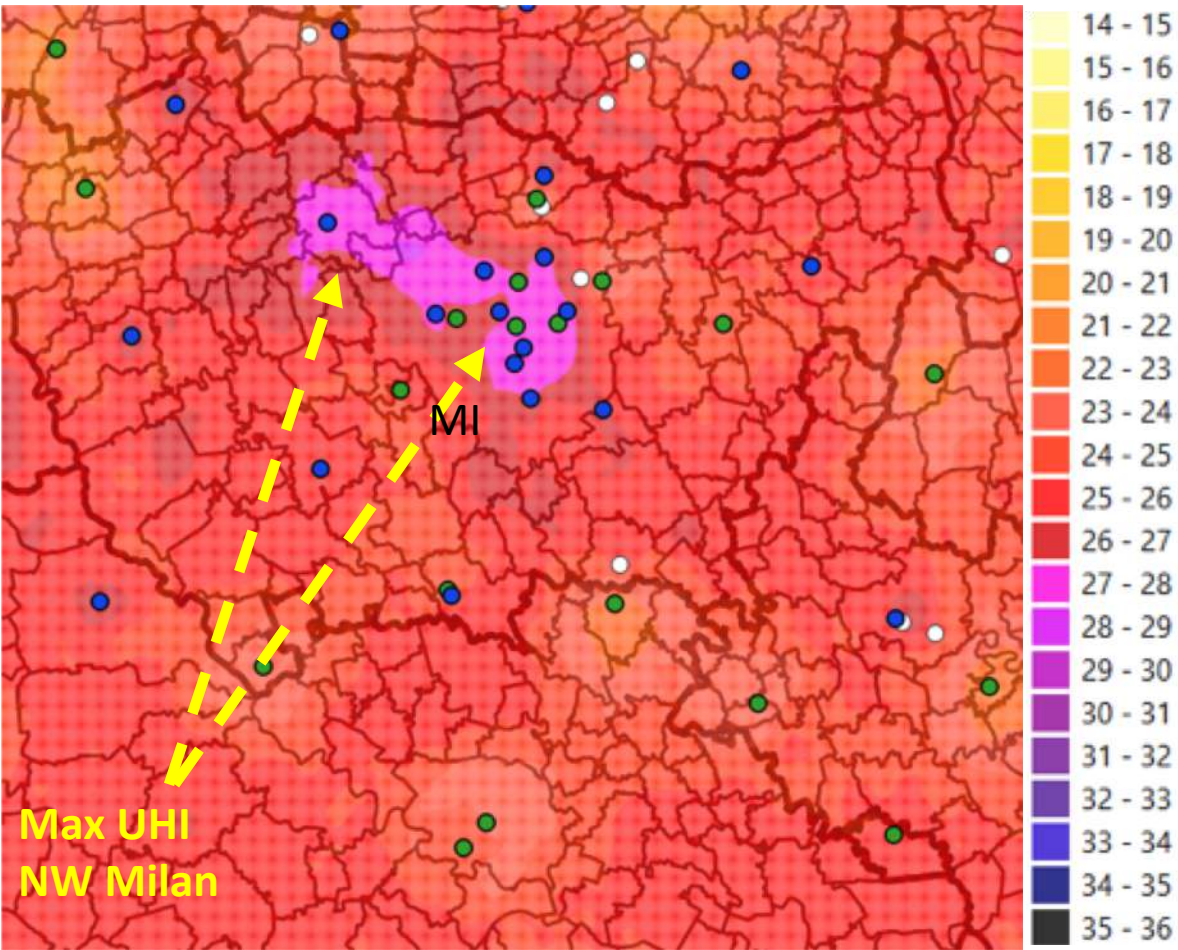
No UHI > 3°C in summer mornings!

Summer UHI 22:00 UTC+1

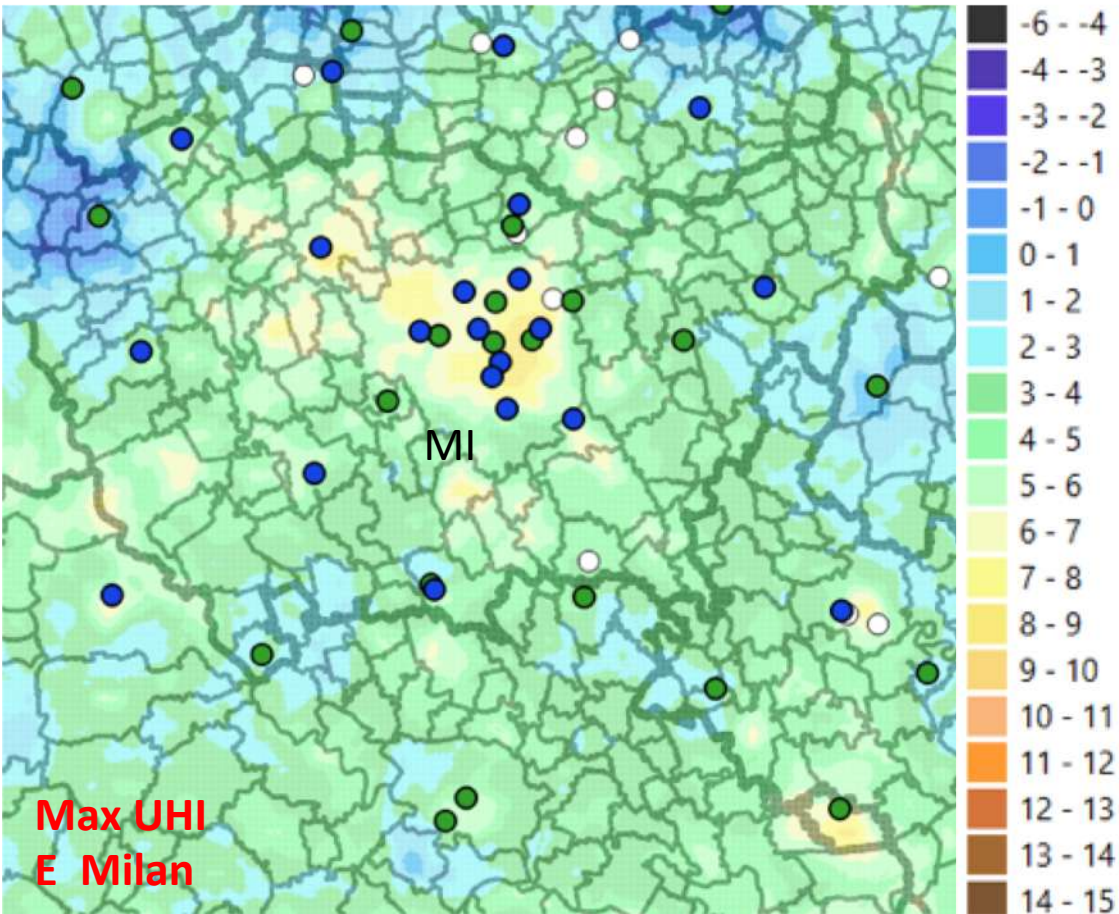


Example: Mean T_{air} in and around Milan at 100 m resolution

Mean T_{air} for STC «Summer_UHI_evening_NW»

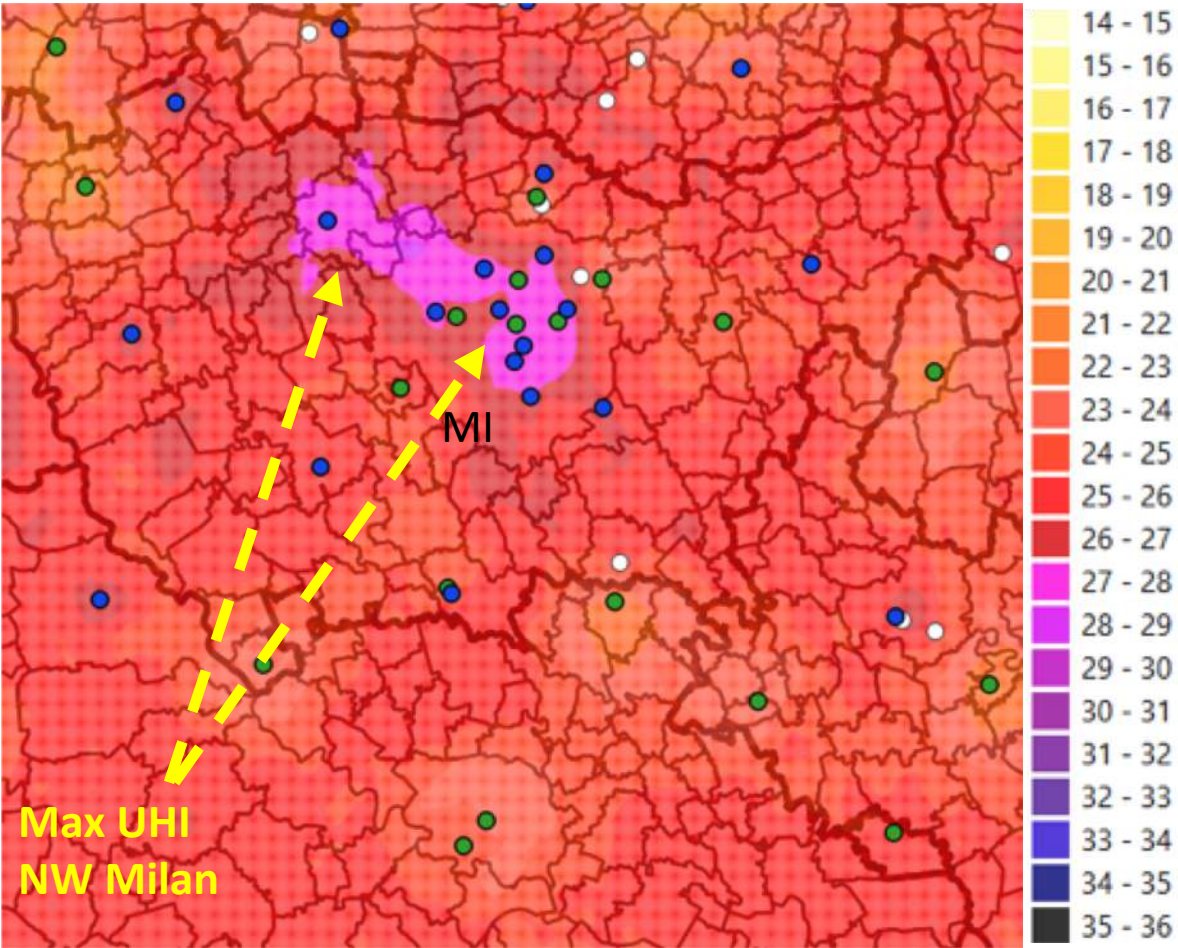


Mean T_{air} for STC «Winter_UHI_evening_E»

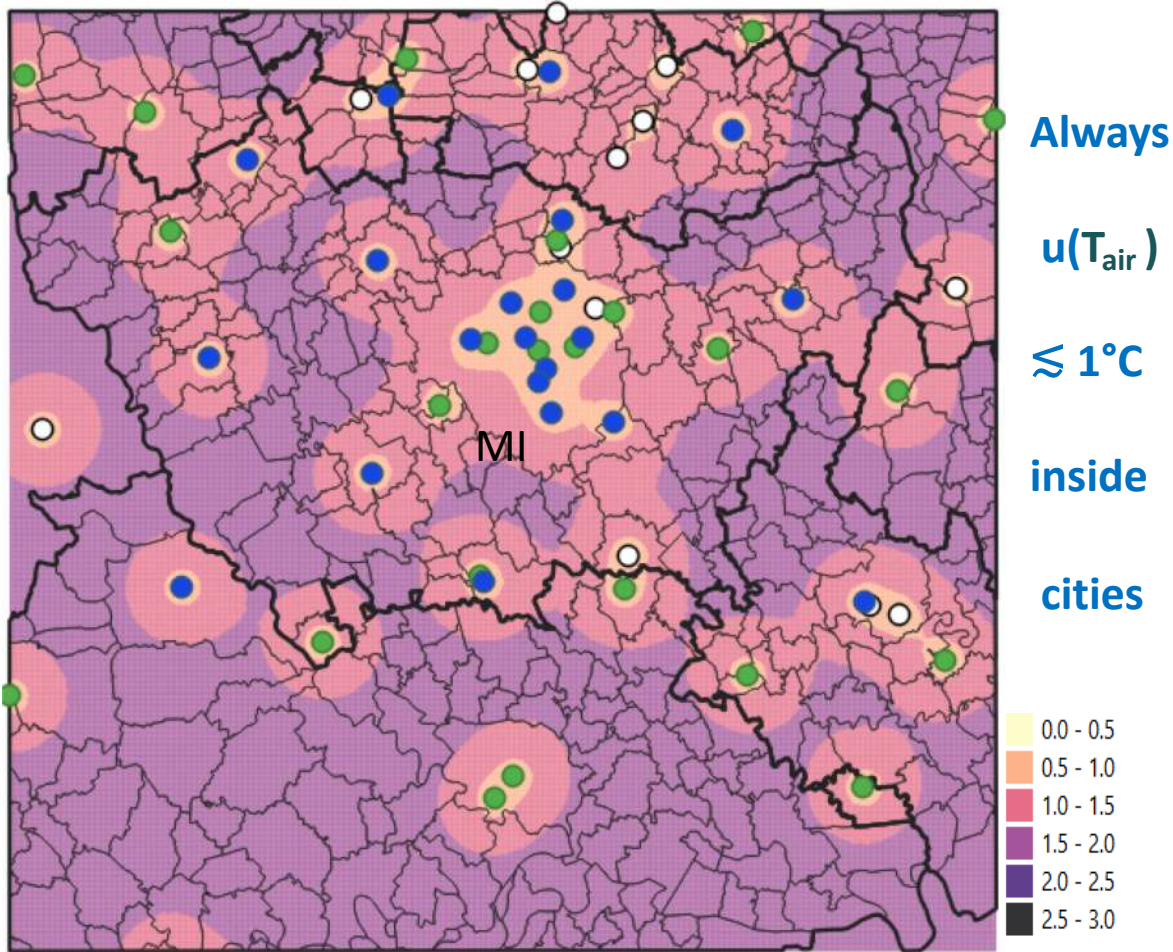


Example: Mean T_{air} uncertainties at 100 m resolution

Mean T_{air} Summer_UHI_NW_22 UTC+1

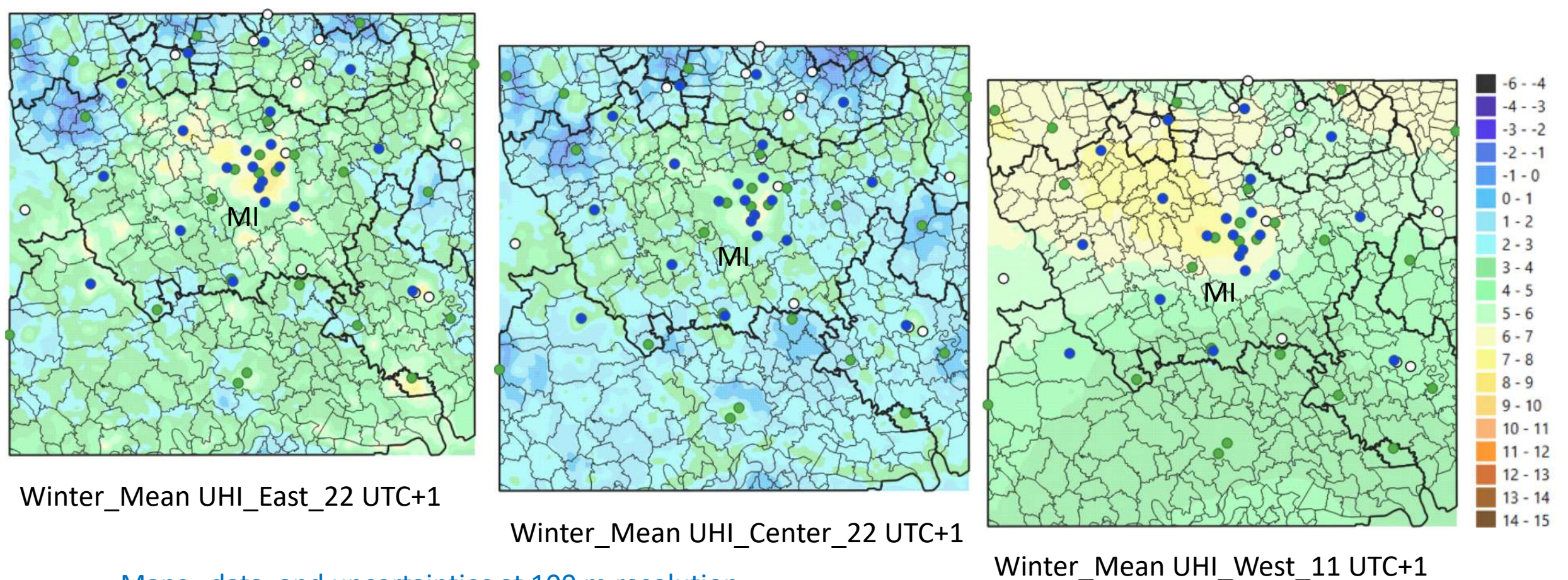


T_{air} uncertainty Summer_UHI_NW_22 UTC+1



Application for adaptation: Milan T_{air} Atlas

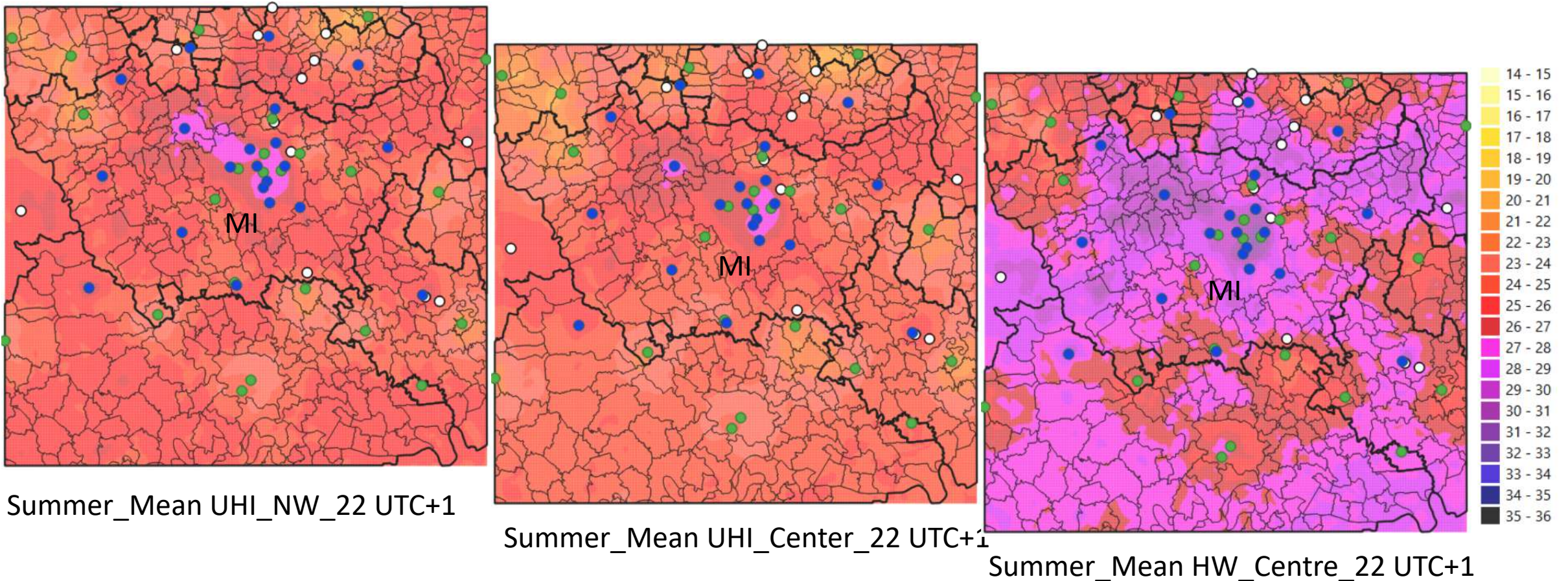
ClimaMi Project: <https://www.progettoclimami.it/atlane> (48 maps)



Maps , data and uncertainties at 100 m resolution

Application for adaptation: Milan T_{air} Atlas

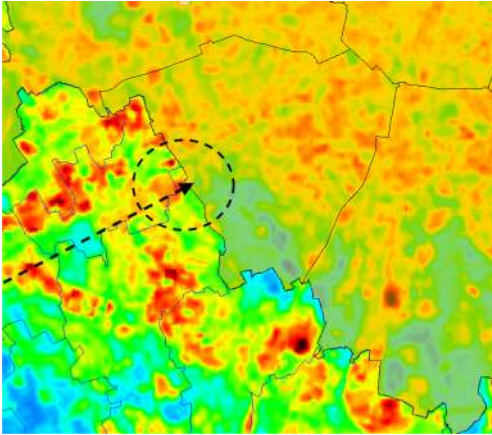
ClimaMi Project: <https://www.progettoclimami.it/atlane> (48 maps)



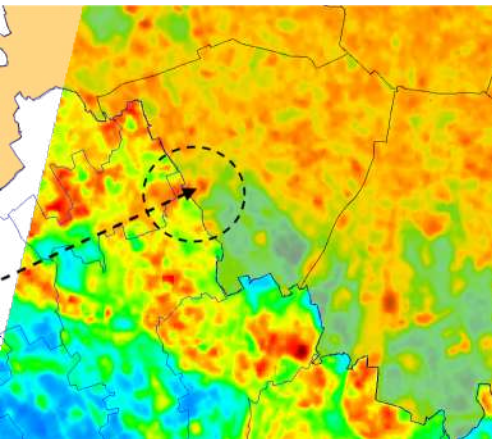
Maps, data and uncertainties at 100 m resolution

LST and COK- T_a differences as effect of an urbanistic variation

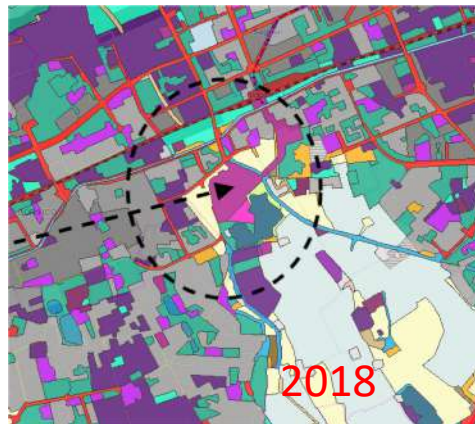
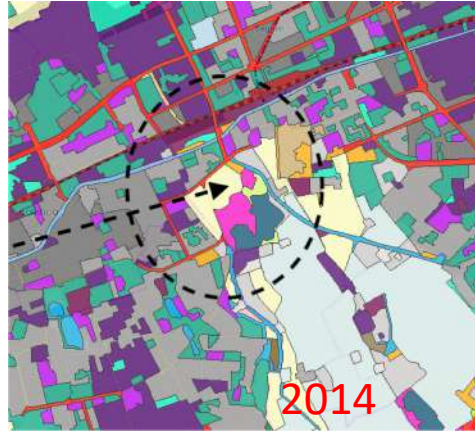
2013-07-16
10:13 UTC



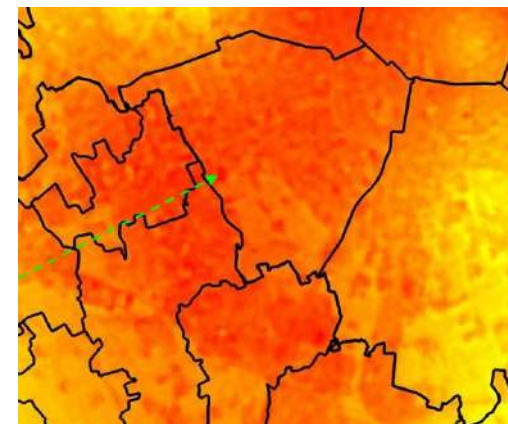
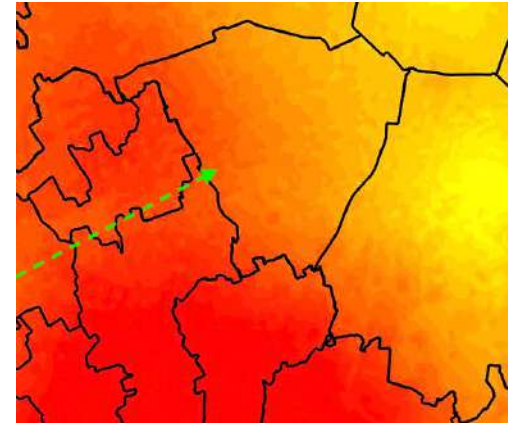
2019-07-17
10:10 UTC



Landsat 8 –LST (30 m)
9 km x 9 km area



Land use – DUSAF
ARPA Lombardia



COK-derived T_a (30 m)
from LST and surface obs.

Ronchetto sul
Naviglio (MI):

SE Milan outskirts,
about 5.7 km from
city centre

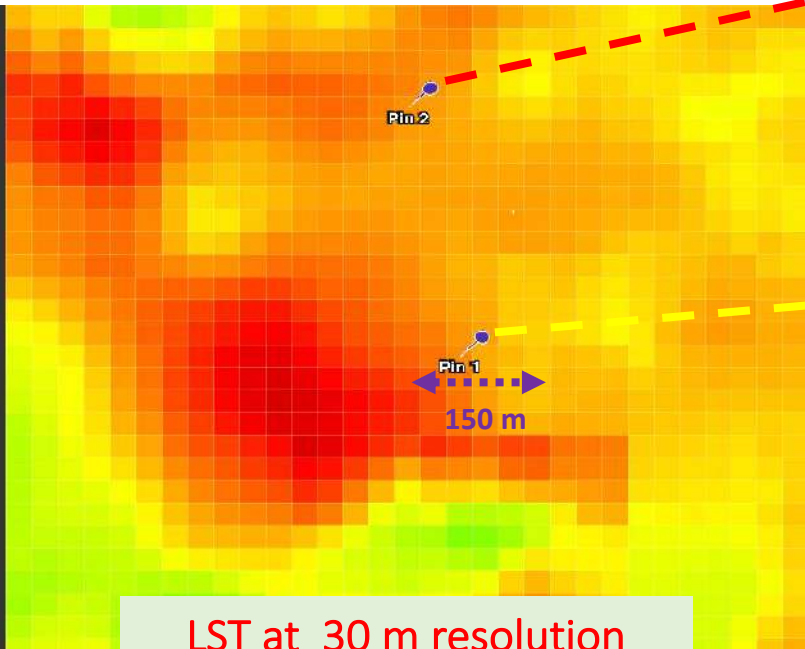
Land use change
from
arable land (2014)
to
industrial
settlement (2018)

Planned: Schiavone Place in Milan at 30 m resolution

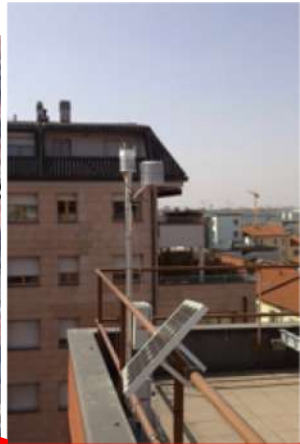
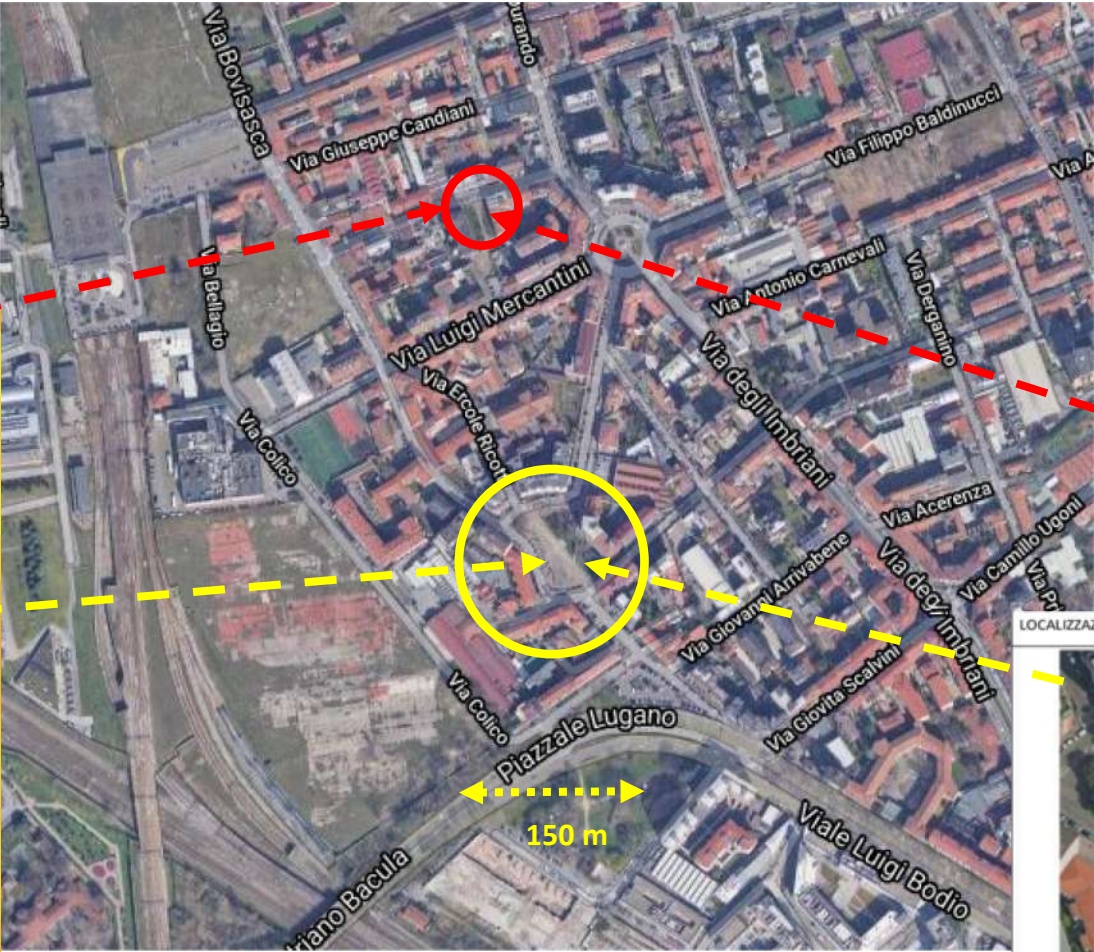
APPLICATION planned:

Assessing effects of adaptation projects in Piazza Schiavone (Milan)

2020-'21



LST at 30 m resolution
2019-07-01 10.11 utc
Piazza Schiavone, Milano



AWS CN04 MI Bovisa,
Measuring T_{air} at top
of UCL



Conclusions

In the framework of the ClimaMi Project (2019-2021)

a **co-kriging methodology** and procedure have been developed and tested making use of:

- high quality air temperature measurements in the UCL by surface operational meteorological networks
- land surface temperatures remotely sensed from space (LST)

to obtain medium- to high-resolution air temperatures fields for climatological as well as

for climate change adaptation purposes (Climatic Services)

Paper submitted for publication in Bull. of Atm. Science and Technology (BAST): High-resolution climatic characterization of air temperature in the Urban Canopy Layer, Montoli • Frustaci • Lavecchia • Pilati