

geoENV  
2018



# CONTRIBUTION OF GEOSTATISTICS IN MAPPING SUBSOIL TEMPERATURE EVOLUTION IN URBAN AREAS

KASMAEE S., TINTI F., BRUNO R.

*Francesco Tinti*

*Department of Civil, Chemical, Environmental and Materials Engineering  
University of Bologna, via Terracini 28, 40131 Bologna Italy*

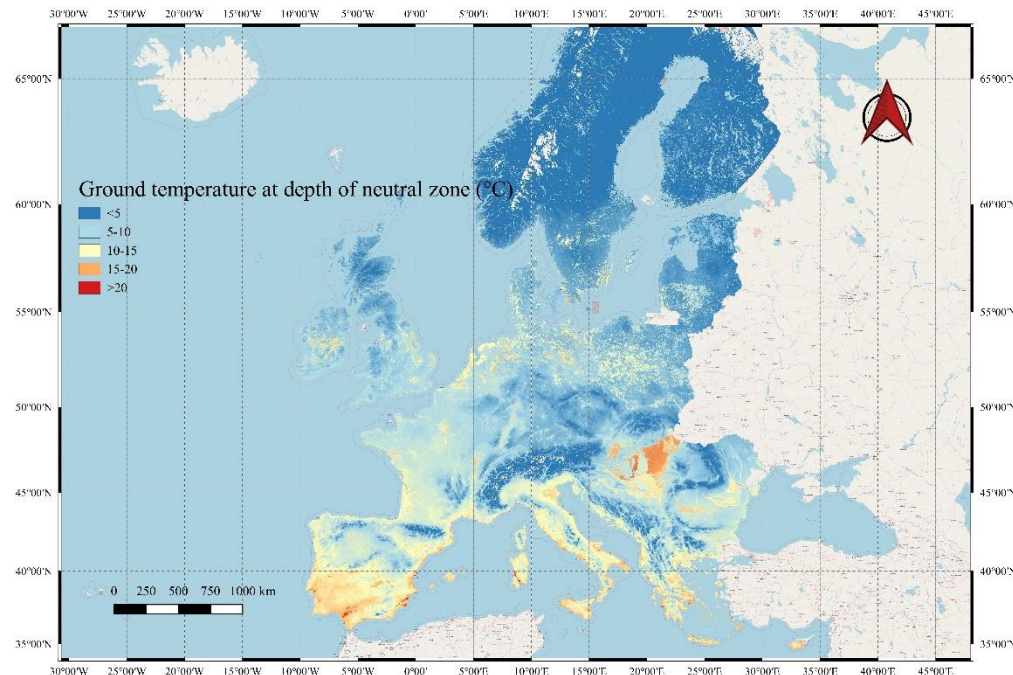
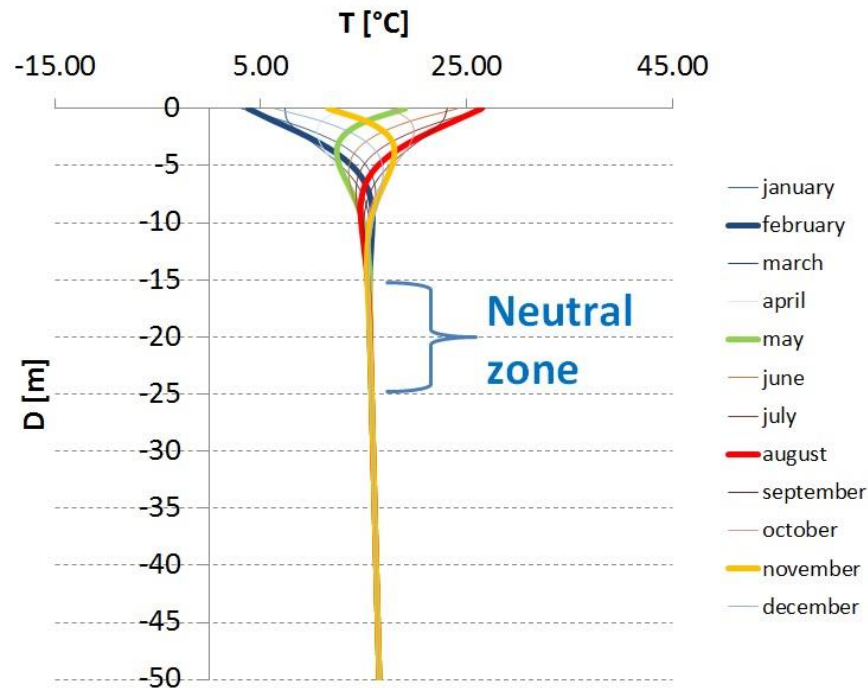
*Belfast, 5<sup>th</sup> july 2018*



# Index

1. The context: ground temperature evolution and its importance for recovery of ambient heat.
2. Deterministic methods and the influence of urbanization.
3. Uncertainty issues and the use of geostatistical simulation.
4. An application of geostatistical simulation on a case study area.
5. Conclusions and next steps.

# 1. The context: ground temperature evolution and its importance for recovery of ambient heat

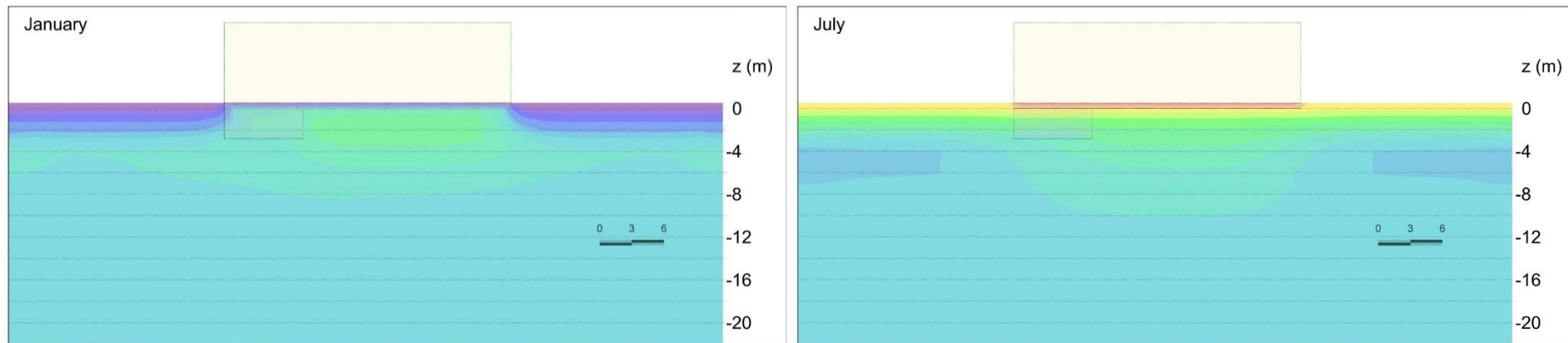


Tinti et al., 2018, *Suitability Evaluation of Specific Shallow Geothermal Technologies Using a GIS-Based Multi Criteria Decision Analysis Implementing the Analytic Hierarchic Process*, *Energies*, 11(2), 457

## ***Recovery of ambient heat:***

Recovery of heat energy at a useful temperature level by means of heat pumps, and which can be stored in the ambient air, **beneath the surface of solid earth (geothermal energy)** or in surface water.

# 1. The context: ground temperature evolution and its importance for recovery of ambient heat



## Key points:

- Temperature variations in space and time at very shallow depth depend on many different factors, such as ground thermal properties, groundwater presence, but also heat losses from single buildings and impact of urban heat island on the ground;
- Ground is a renewable reservoir of ambient heat, with many applications.

**Objective** → Improvement of design of ambient heat recovery systems through a more accurate estimation of ground temperature evolution.

## 2. Deterministic methods and the influence of urbanization



### Underground thermal properties

$\vec{\lambda}_g(\tilde{x})$  = ground thermal conductivity  $\left[\frac{W}{mK}\right]$

$c_g(\tilde{x})$  = ground volumetric heat capacity  $\left[\frac{J}{m^3K}\right]$

$\vec{\alpha}_g(\tilde{x}) = \frac{\vec{\lambda}_g}{c_g}$  = ground thermal diffusivity  $\left[\frac{m^2}{s}\right]$

Ground thermal diffusivity determines the dampening of climatic wave into the ground. It varies for different layers.

Ground thermal conductivity influences the impact of geothermal heat flow density on the ground temperature. It varies for different layers.

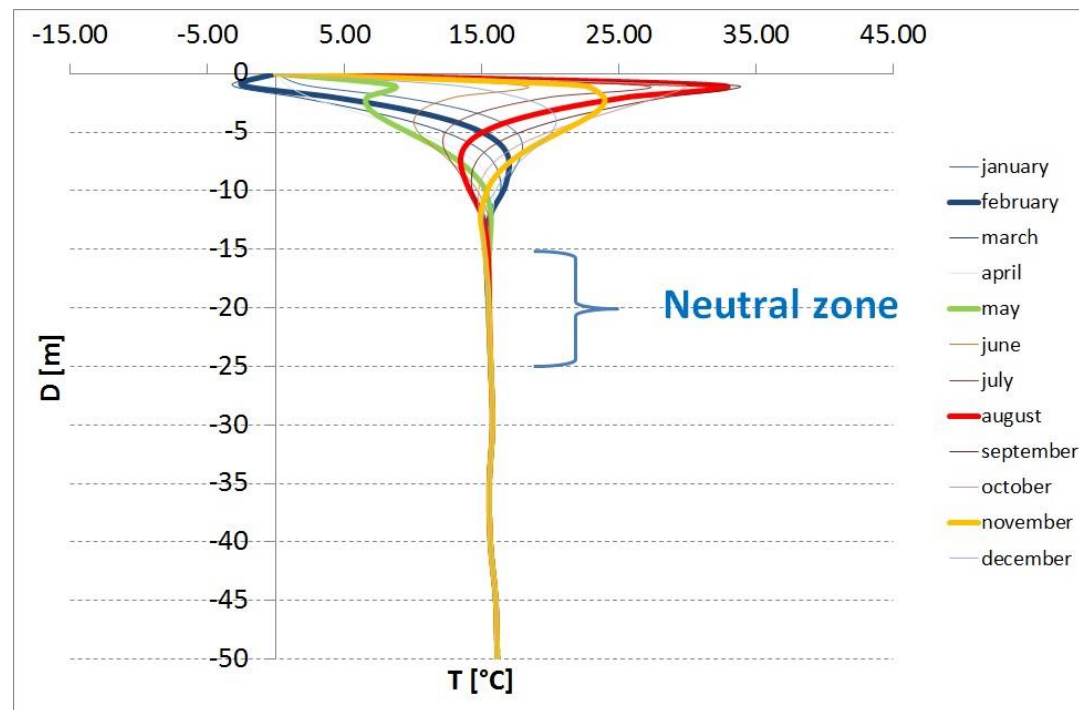


## 2. Deterministic methods and the influence of urbanization

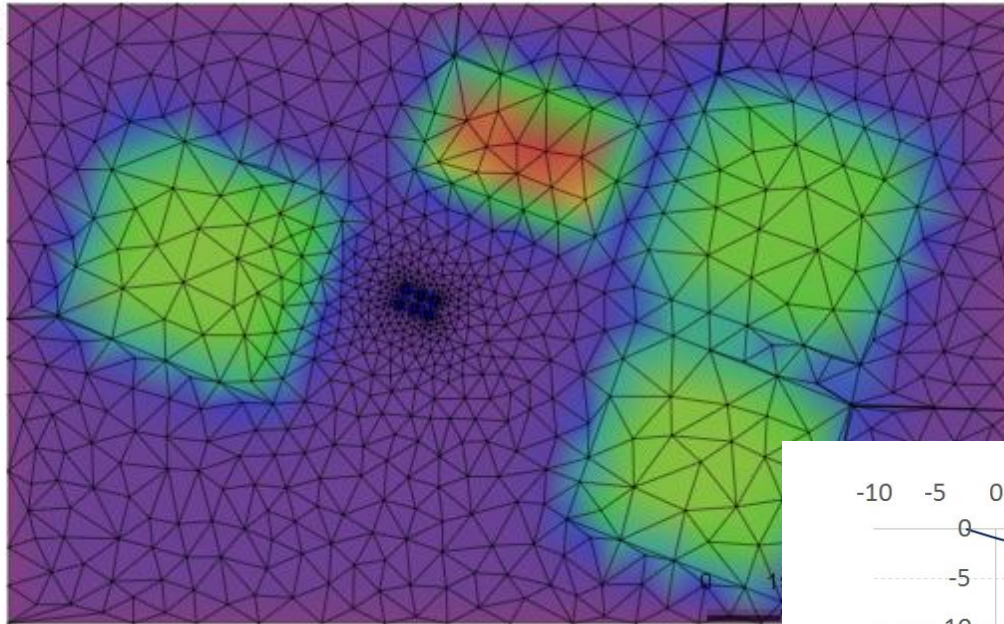
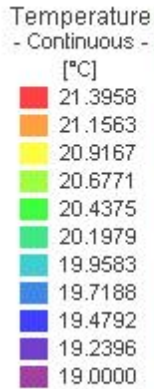
# Underground temperature distribution assessment

$$T_g(d, t) = T_m - A \cdot \exp\left[-d \cdot \sqrt{\left(\frac{\pi}{T \cdot \alpha_g}\right)}\right] \cdot \cos\left[\frac{2\pi}{T} \cdot \left(t - t_{T_0} - \frac{d}{2} \cdot \sqrt{\left(\frac{T}{\pi \cdot \alpha_g}\right)}\right)\right] + \vec{\nabla} T_{geo}(h, \lambda_g) \cdot d$$

- $T_g$  → Underground temperature (°C)  
 $T_m$  → Annual average temperature (°C)  
 $A$  → Wave amplitude (°C)  
 $T$  → Wave period (d)  
 $t_{T_0}$  → Day of minimum temperature (days)  
 $\alpha_g$  → Ground thermal diffusivity (m<sup>2</sup>/days)  
 $\lambda_g$  → Ground thermal conductivity (W/(mK))  
 $C_g$  → Ground heat capacity (J/(kgK))  
 $\rho_g$  → Ground density (kg/m<sup>3</sup>)  
 $d$  → Depth (m)  
 $\vec{\nabla} T_{geo}$  → Geothermal gradient (°C/m)  
 $h$  → Geothermal heat flow (W/m<sup>2</sup>)



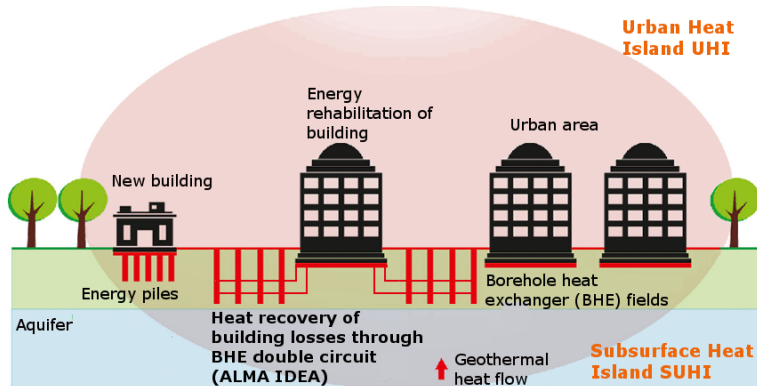
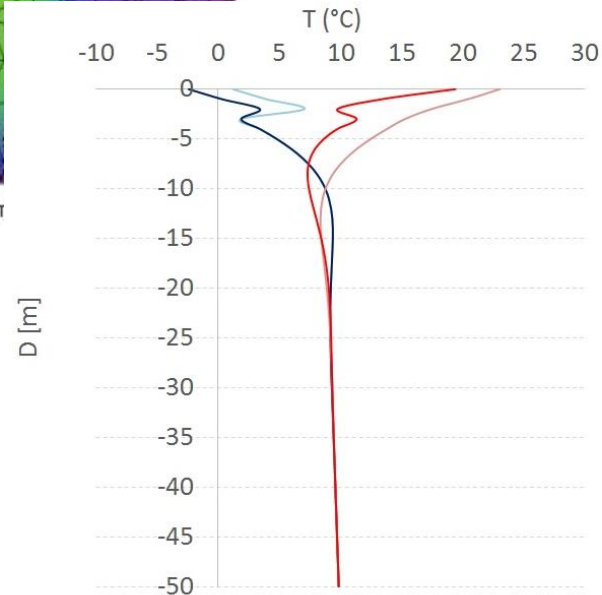
## 2. Deterministic methods and the influence of urbanization



FEFLOW (R)

0 [d]

[m]



## 2. Deterministic methods and the influence of urbanization



City	Area (km <sup>2</sup> )	Population density (inhab./km <sup>2</sup> )	Aquifer geology	Thickness (m)	Porosity (-)	Potential heat content (kJ/m <sup>2</sup> )	Heating demand (kJ/year)
Cologne	405	2528	gravel, sand	15-30	0.15-0.25	4.8×10 <sup>10</sup> - 4.8×10 <sup>11</sup>	1.9×10 <sup>10</sup>
Winnipeg	5302	1429	carbonate	5-15	0.05-0.10	2.2×10 <sup>10</sup> - 2.1×10 <sup>11</sup>	4.1×10 <sup>10</sup>
Shanghai	6200	2646	sand, clay	10-20	0.20-0.30	5.0×10 <sup>10</sup> - 3.5×10 <sup>11</sup>	2.3×10 <sup>9</sup>
Tokyo	2187	5874	sand, clay	30-70	0.20-0.30	5.0×10 <sup>10</sup> - 7.0×10 <sup>11</sup>	2.5×10 <sup>10</sup>
London	1707	4761	chalk	30-40	0.05-0.20	1.1×10 <sup>11</sup> - 5.6×10 <sup>11</sup>	9.5×10 <sup>10</sup>
Istanbul	1830	6211	limestone	10-30	0.05-0.25	4.4×10 <sup>10</sup> - 5.0×10 <sup>11</sup>	5.5×10 <sup>9</sup>
Prague	496	2504	sandstone	10-30	0.10-0.30	4.6×10 <sup>10</sup> - 5.3×10 <sup>11</sup>	9.6×10 <sup>9r</sup>

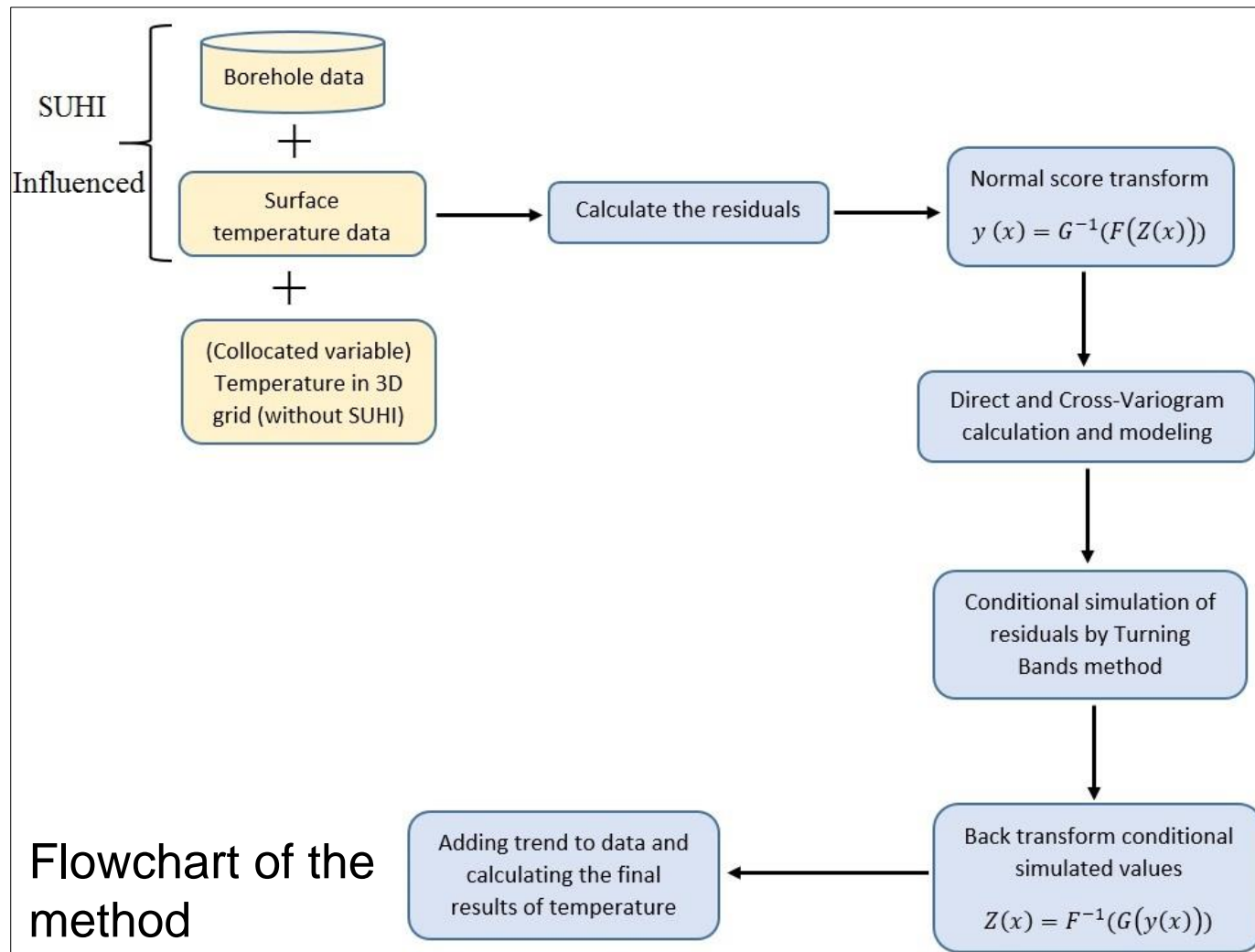
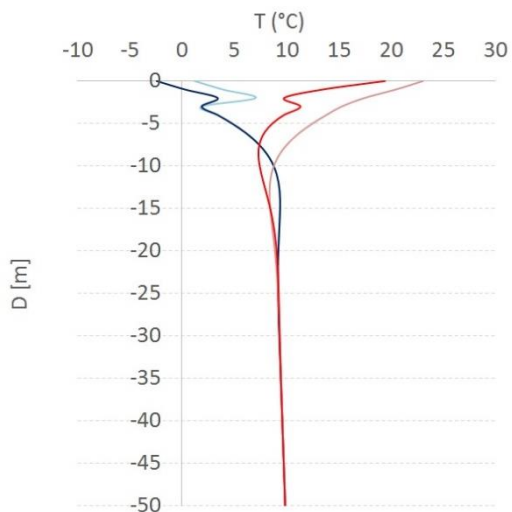
*Zhu K., 2013 Urban Heat Island in the Subsurface and Geothermal Potential in Urban Areas PhD Thesis, University of Tübingen, Germany*

- Deterministic calculation based on climate data and ground properties
- Statistical correlations between population density and temperature increase



### 3. Uncertainty issues and the use of geostatistical simulation

## Geostatistics applied to the topic





### 3. Uncertainty issues and the use of geostatistical simulation



#### Choice of an area of intervention

#### Data to get

Temperature direct measurements – Direct variable

Borehole temperature data (measured)

Surface temperature data (measured)

Values affected by the presence of urbanization

3D temperature calculation – Collocated variable

Deterministic calculation of ground temperature in 3D from estimated parameters and indicators at macro scale (geological layers, thermal conductivity, thermal diffusivity, climate data, heat flow density)



### 3. Uncertainty issues and the use of geostatistical simulation



#### Procedure for geostatistical analysis of data (1/2)

Removal of the trend in the vertical direction and work on the residuals as fluctuation of the variables: target variable (T meas) and collocated variable

Normal score transform of residuals

$$y(x) = G^{-1} \left( F \left( Z(x) \right) \right)$$

Variograms on the residual of the target variable

Variograms on the residual of the collocated variable (T det. calculation)

Cross-variograms between the residual of the target variable and the residual collocated variable

### 3. Uncertainty issues and the use of geostatistical simulation



#### Procedure for geostatistical analysis of data (2/2)

Conditional simulation of residuals of direct variable by Turning Bands method

Back transform of conditional simulation results  $Z(x) = F^{-1}(G(y(x)))$

Addition of trend to results

Calculation of final 3D temperature in the area of interest, after considering the effective temperature measurements, affected by presence of urbanization





### 3. An application of geostatistical simulation on a case study area



Area of interest:  
Zurich City Area



Information from:  
 Bayer et al., 2016  
*Extracting past atmospheric warming and urban heating effects from borehole temperature profiles*,  
 GEOTHERMICS, 64  
 289-299

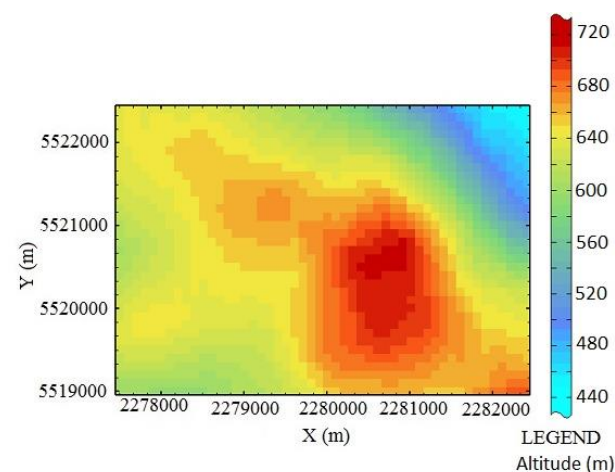


### 3. An application of geostatistical simulation on a case study area

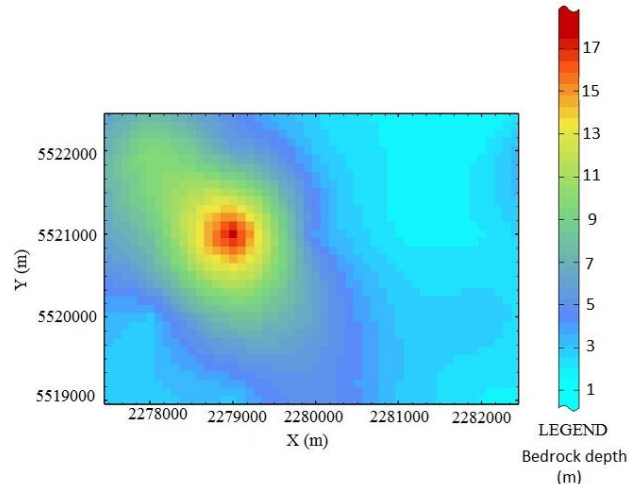
#### Area of interest: Zurich City Area

#### Reasons of area selection:

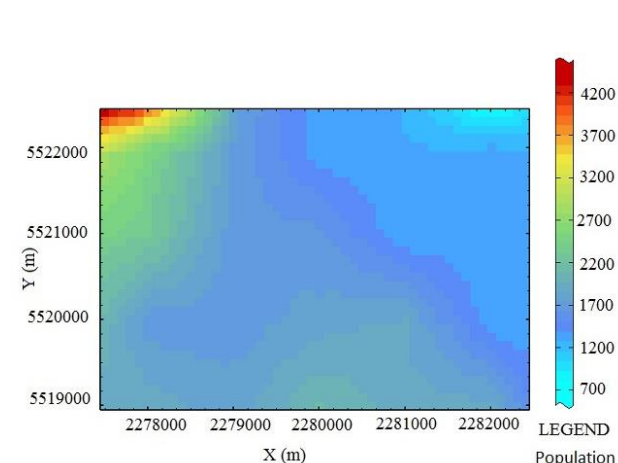
- No presence of groundwater bodies;
- High geological variability: presence of valleys and mountains;
- High population density variability: presence of urban (villages) and rural areas.



Altitude  
m a.s.l.

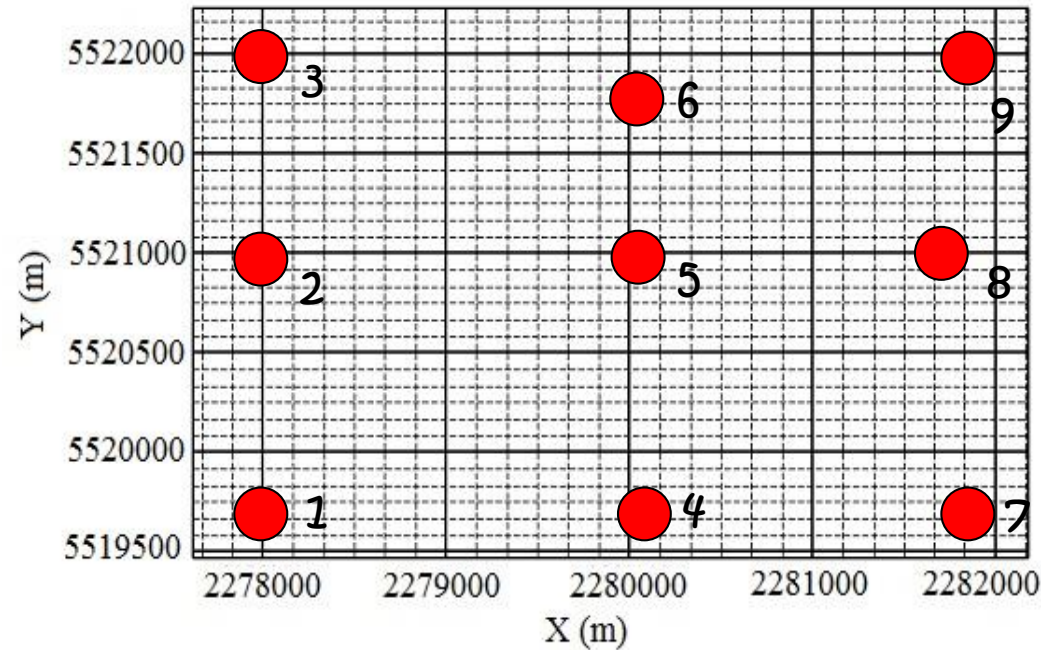


Bedrock depth  
m b.g.l.

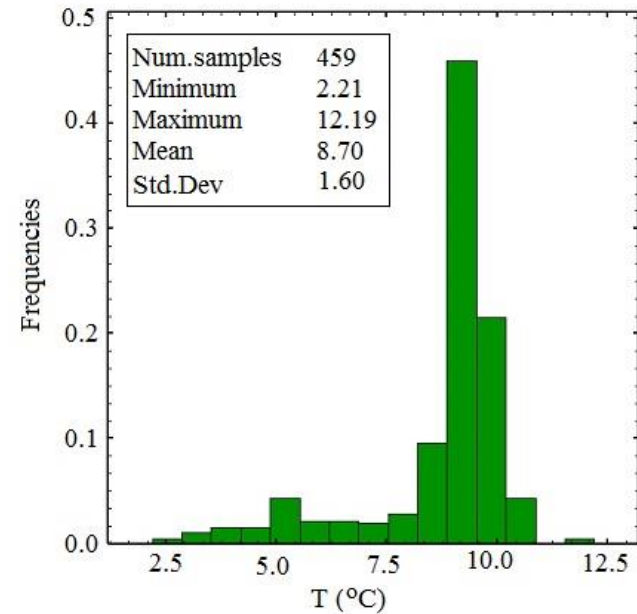


Population density  
Inhab./km<sup>2</sup>

### 3. An application of geostatistical simulation on a case study area

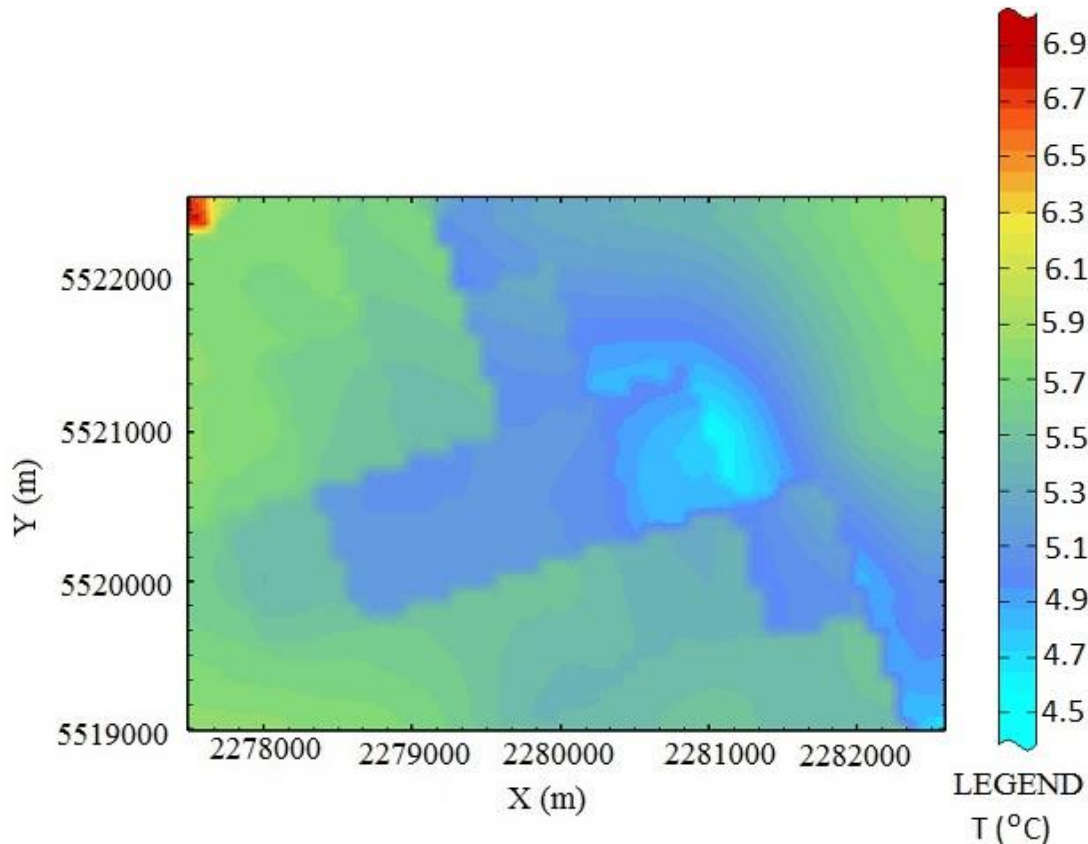


Location of 9 artificial boreholes in the study area, with CRS: ED50 / UTM Zone 28



Histogram of temperature data and related basic statistics

### 3. An application of geostatistical simulation on a case study area



Surface temperature in the area of interest for March

The value integrates climate information with urban effect

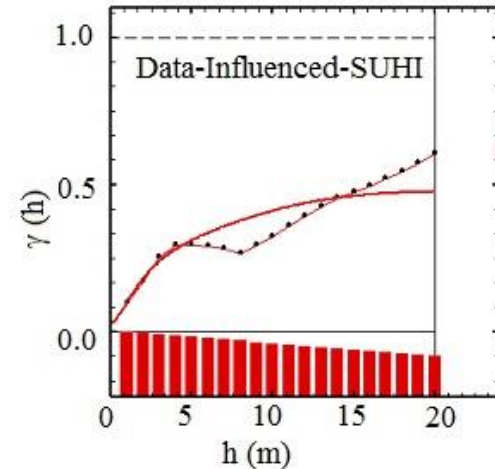
### 3. An application of geostatistical simulation on a case study area



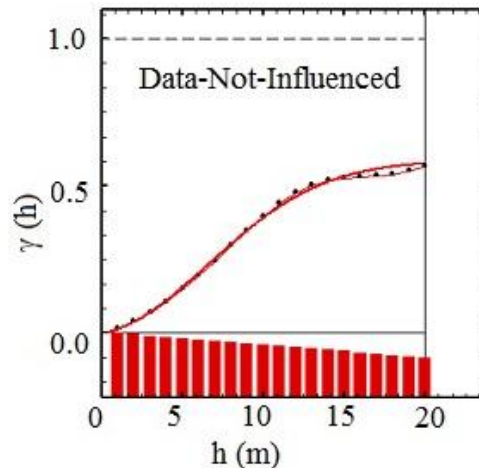
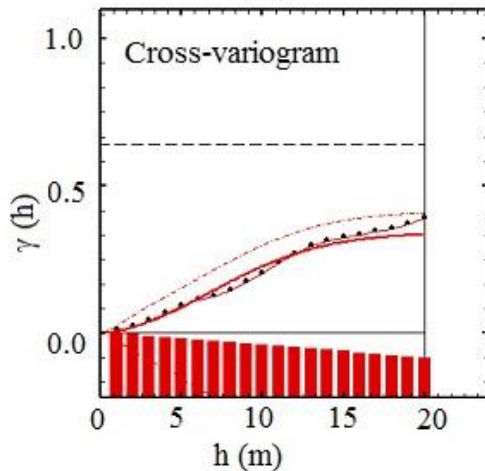
#### Application on data

- Removing trend  $Z(x) = m(x) + Y(x)$
- Normal score transformation with Hermit Polynomials  $y(x) = G^{-1}(F(Z(x)))$

### 3. An application of geostatistical simulation on a case study area



Direct and cross-variograms in vertical direction (sample and models)



$$\left\{ \begin{array}{l} \gamma_{K-Bessel}(h) = C \cdot \left[ 1 - \frac{\left(\frac{h}{a}\right)^k}{2^{k-1} \cdot \Gamma(\alpha)} K_{-k}\left(\frac{h}{a}\right) \right] \\ k > 0 \end{array} \right.$$

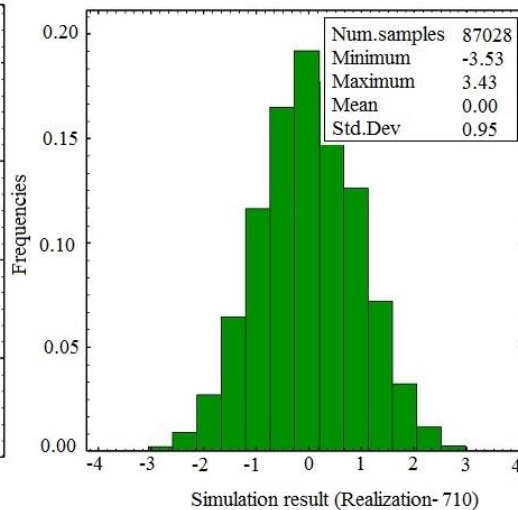
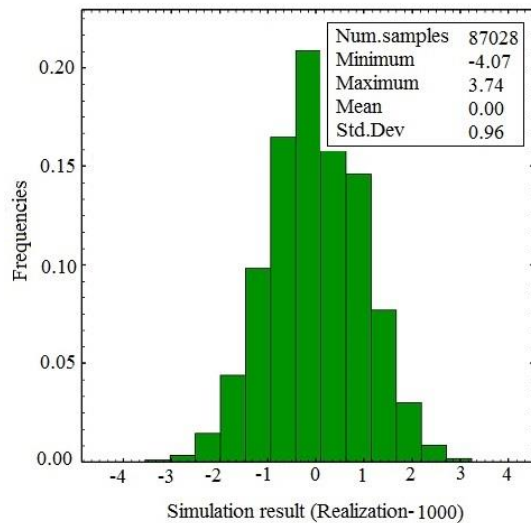
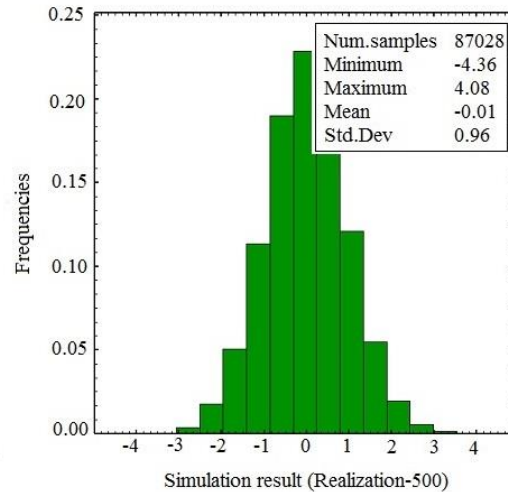
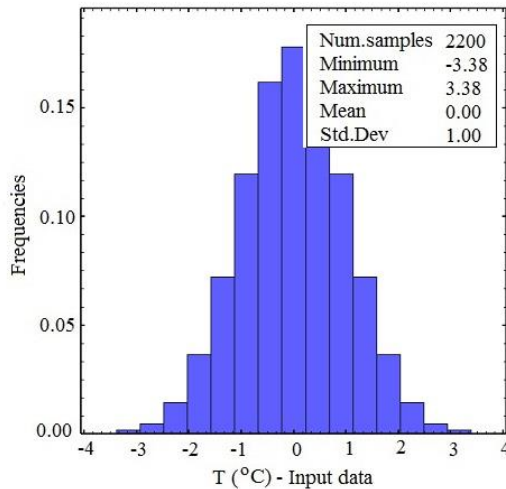
$$\left\{ \begin{array}{l} \gamma_{J-Bessel}(h) = C \cdot \left[ 1 - 2^k \cdot \Gamma(\alpha + 1) \frac{J_k\left(\frac{h}{a}\right)}{\left(\frac{h}{a}\right)^k} \right] \\ k > \frac{sd}{2} - 1 \end{array} \right.$$

Variogram model of ground T from:

Kasmaee et al., *Use of Universal Kriging as a tool to estimate mountain temperature distribution affected by underground infrastructures: the case of the Brenner Base Tunnel*, in: European Geothermal Congress 2016 Proceedings, Bruxelles



### 3. An application of geostatistical simulation on a case study area



#### Conditional simulation with turning bands

Number of turning bands: 400

Number of simulations: 1000

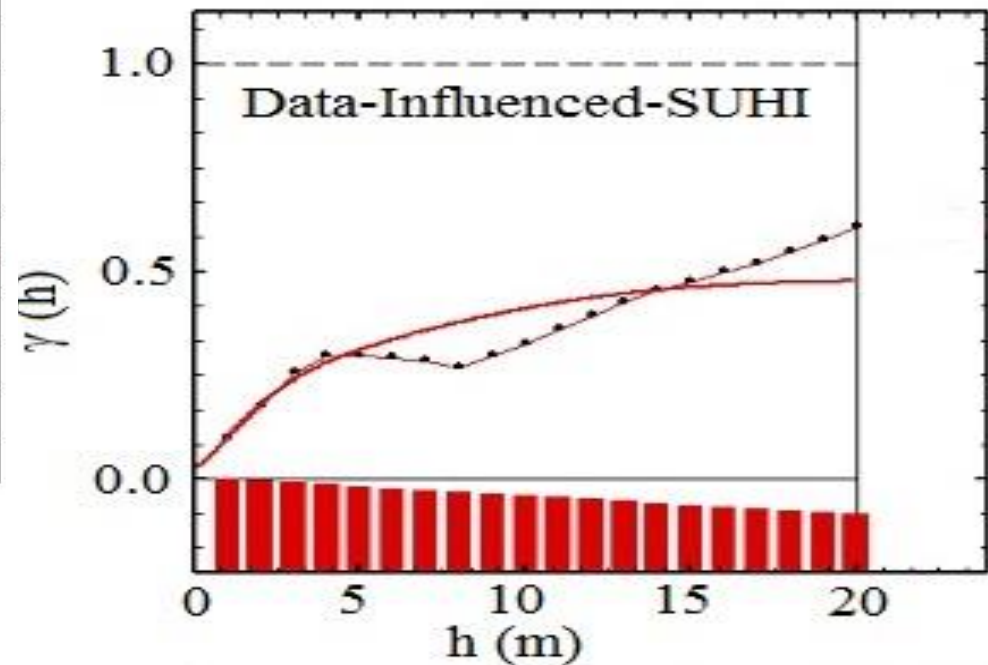
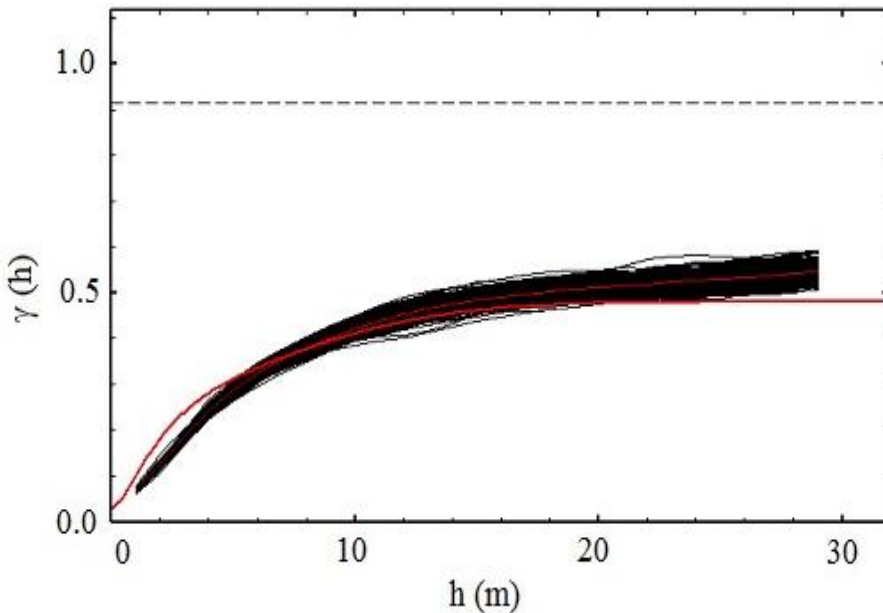
Implementation: ISATIS

Simulation time: 3 hours

Comparisons of input data (Gaussian distribution) with the distributions of simulation results for three realizations.

### 3. An application of geostatistical simulation on a case study area

Convergence of the variogram results obtained from all simulation realizations with the initial variogram model used for the simulation (red thick line)



Comparison between the variogram of simulation realisation and the variogram model of residuals of initial temperature data influenced by urban presence





### 3. An application of geostatistical simulation on a case study area

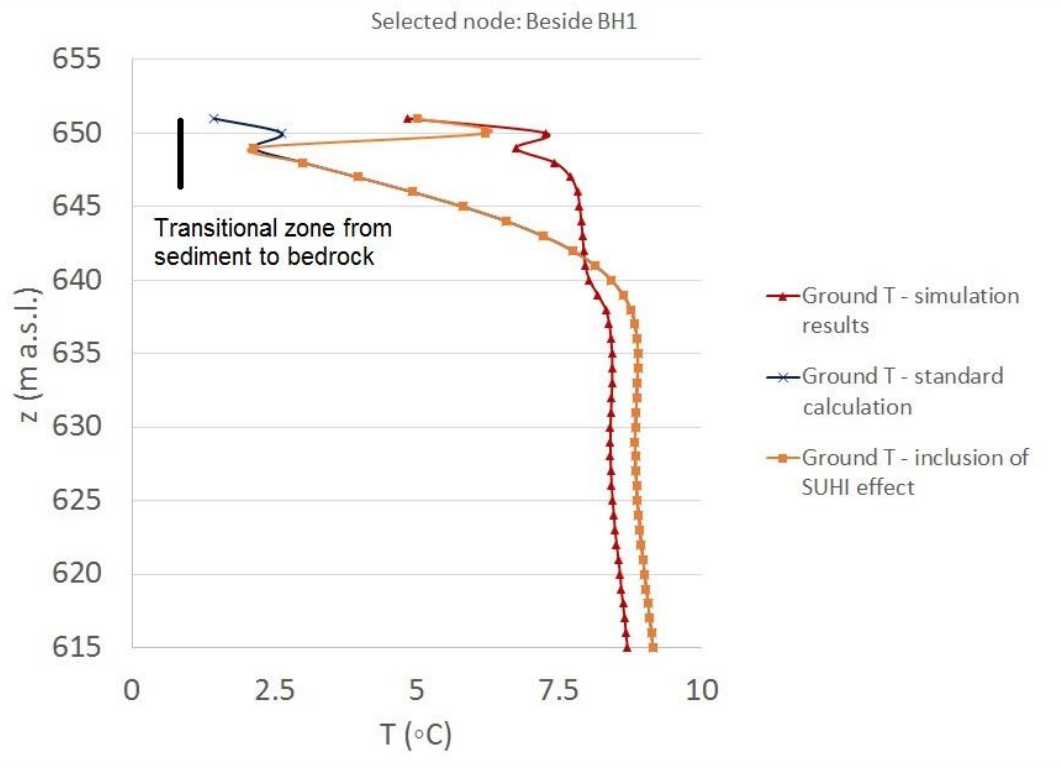


Ex.  
Realisation 1  
Beside BH1

◆ Borehole temperature profiles measurements (from Bayer et al., 2016)  
★ Selection of grid nodes results of simulation



### 3. An application of geostatistical simulation on a case study area







### 3. An application of geostatistical simulation on a case study area



◆ Borehole temperature profiles measurements (from Bayer et al., 2016)

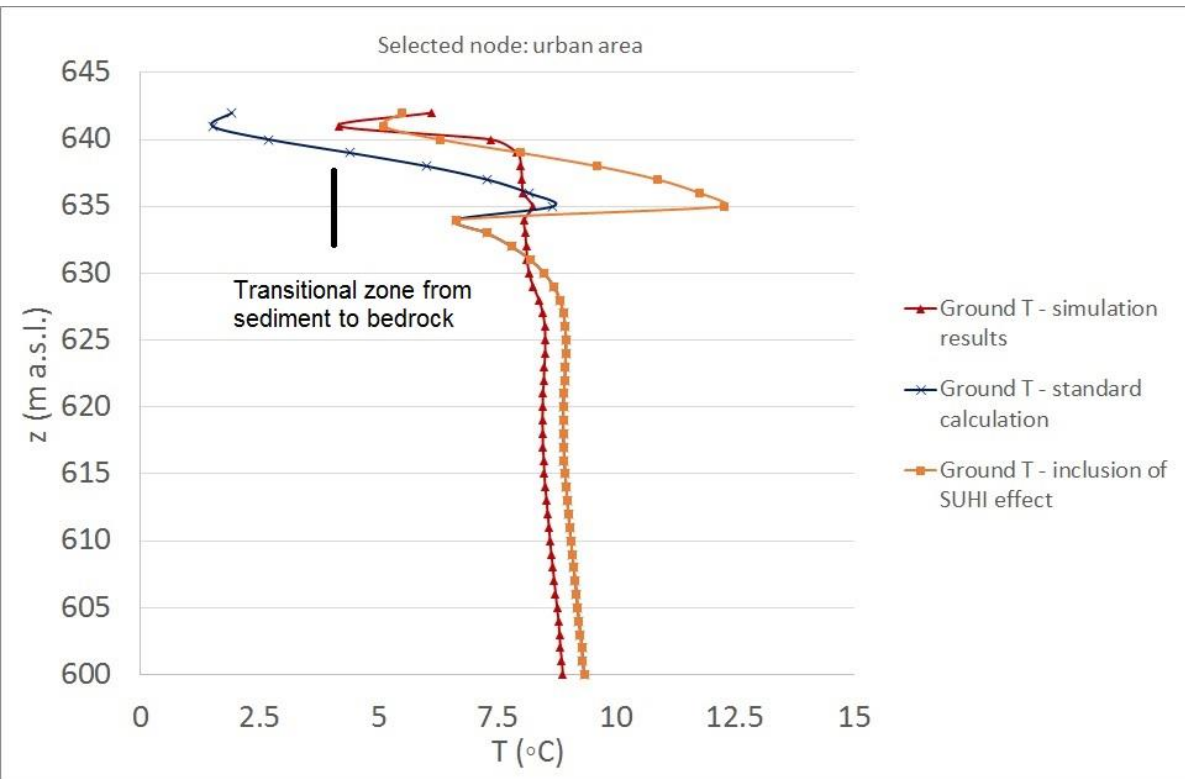
★ Selection of grid nodes results of simulation

Ex.  
Realisation 2  
In urban area





### 3. An application of geostatistical simulation on a case study area







### 3. An application of geostatistical simulation on a case study area

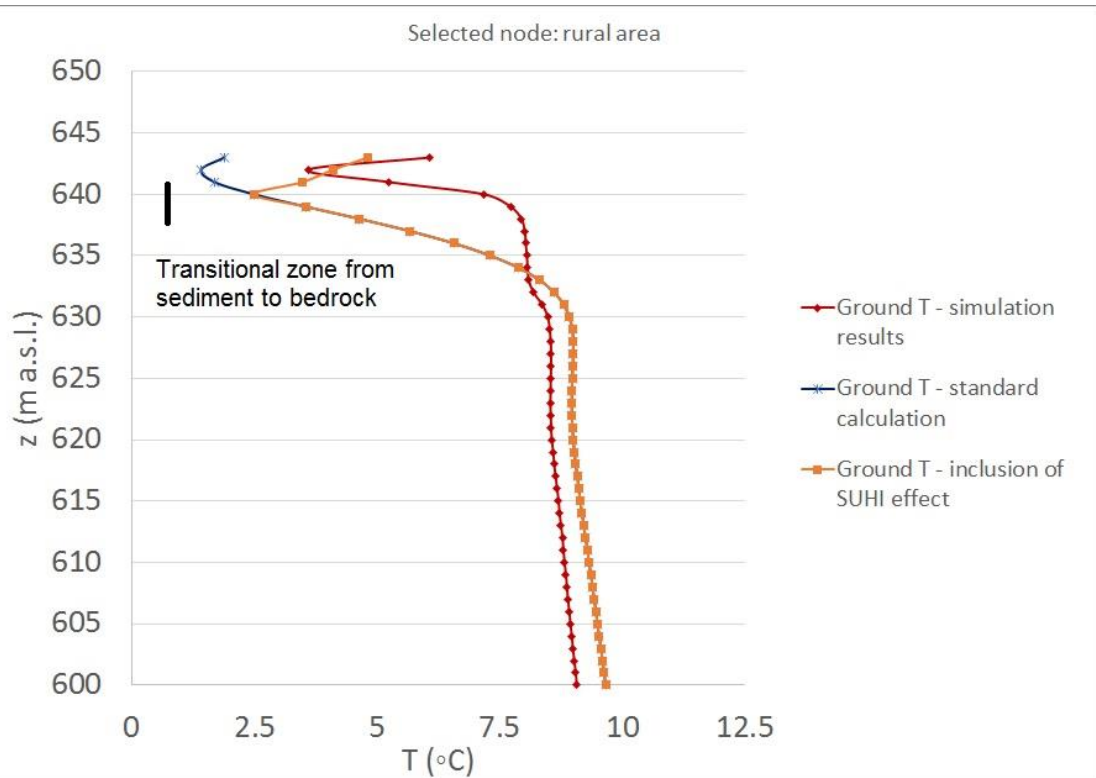


Ex.  
Realisation 2  
  
Out of urban  
area





### 3. An application of geostatistical simulation on a case study area





## 4. Conclusions and next steps



### Preliminary results

- Geostatistical conditional simulation can be used to reconstruct in 3D ground temperature, including urban heat island (UHI) effect.
- Sample data are surface temperature, including UHI effect (for example aerial thermography), and temperature measurements in depth at convenient scale (measurements from existing wells, boreholes,...).

### Impact for geothermal sector

- Improved characterization of geothermal resource, for its sustainable exploitation at shallow depth (< 50 m) with geothermal heat exchangers





## 4. Conclusions and next steps



# Next steps and future work

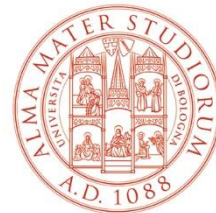
- Adding the fourth dimension (time) in the simulation.
- Inclusion of geology information as indicator (with transition between sediments and bedrock).
- Validation of the methodology on a real case with borehole temperature measurements in a defined future area of Bologna (IT), with aerial thermography and borehole and well samples.
- Inclusion of additional temperature measurements from city underground structures and spaces (cellars, garages, sewer system, channels).



CONTRIBUTION OF GEOSTATISTICS IN MAPPING SUBSOIL  
TEMPERATURE EVOLUTION IN URBAN AREAS



# Thank you for your attention



ALMA MATER STUDIORUM  
UNIVERSITÀ DI BOLOGNA

Sara Kasmaee Yazdi, Francesco Tinti, Roberto Bruno

DICAM - *Department of Civil, Chemical, Environmental and Materials Engineering*  
*via Terracini 28, 40131 Bologna, Italy*

[francesco.tinti@unibo.it](mailto:francesco.tinti@unibo.it)

[www.dicam.unibo.it](http://www.dicam.unibo.it)

The research work presented was supported by the research project GEOTeCH ([www.geotech-project.eu](http://www.geotech-project.eu)), co-funded by the European Community Horizon 2020 Program for European Research and Technological Development (2014–2020)—Grant Agreement 656889.