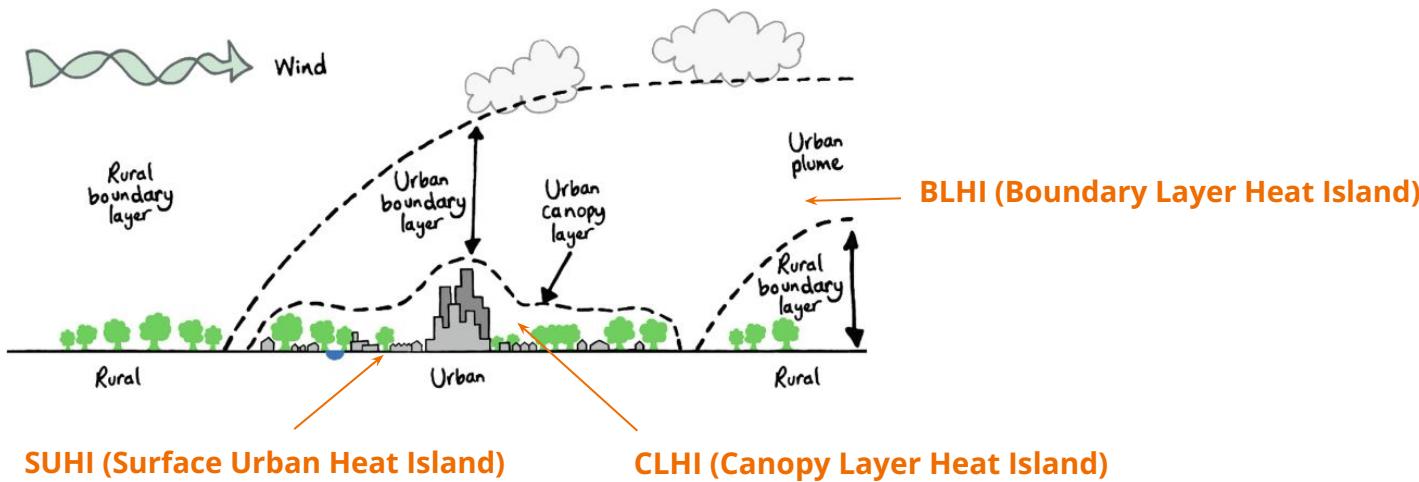

Refinement of Urban Heat Island mapping using Landsat satellite imagery

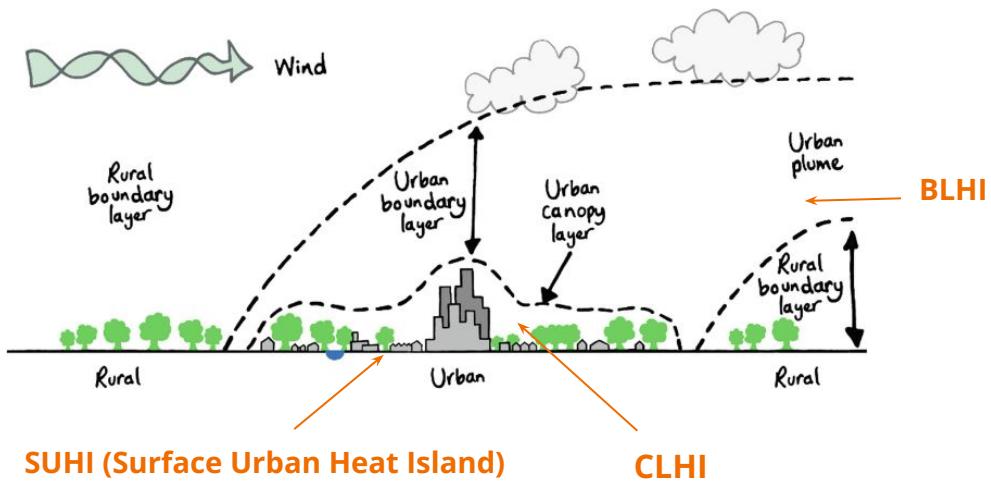
Hana Bobáľová

Department of Physical Geography and Geoinformatics
Comenius University in Bratislava

Urban Heat Island (UHI)



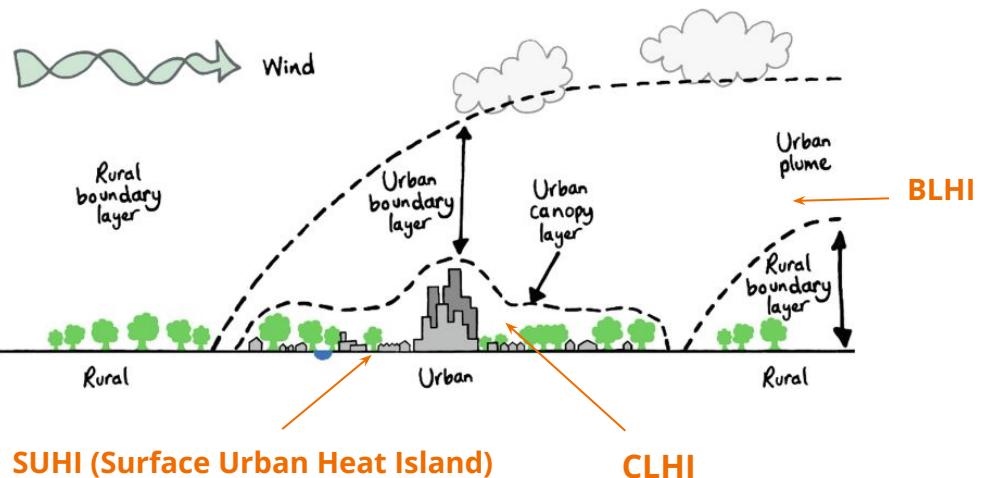
Urban Heat Island (UHI)



Satellite images in Thermal Infrared (TIR) band of the spectrum

Landsat (TM, ETM, TIRS), MODIS, ASTER, VIIRS, Sentinel-3, Ecostress ...

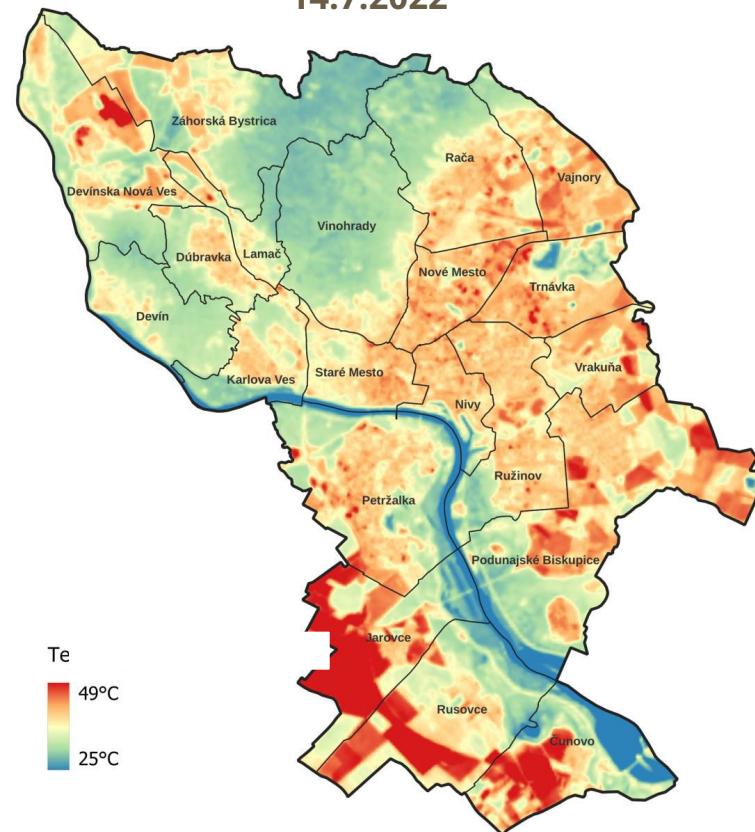
Urban Heat Island (UHI)



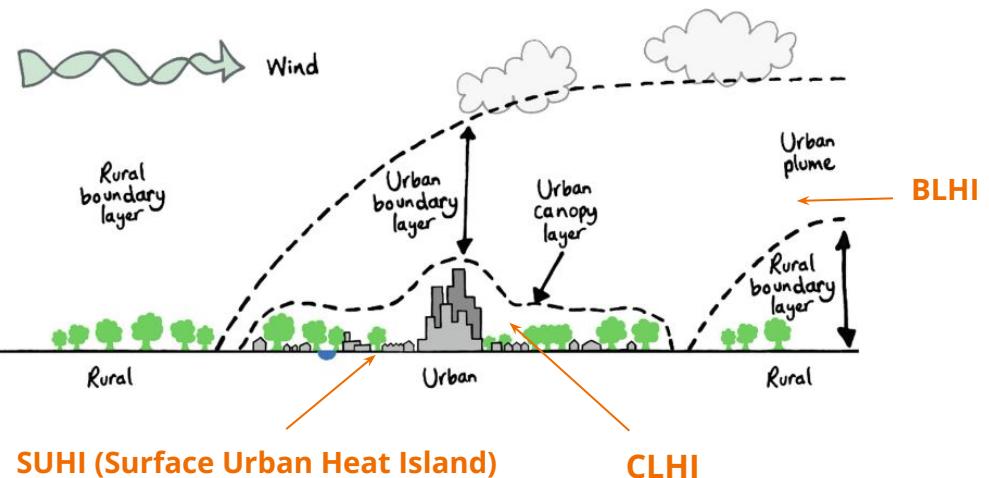
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Surface Temperature
in Bratislava
14.7.2022



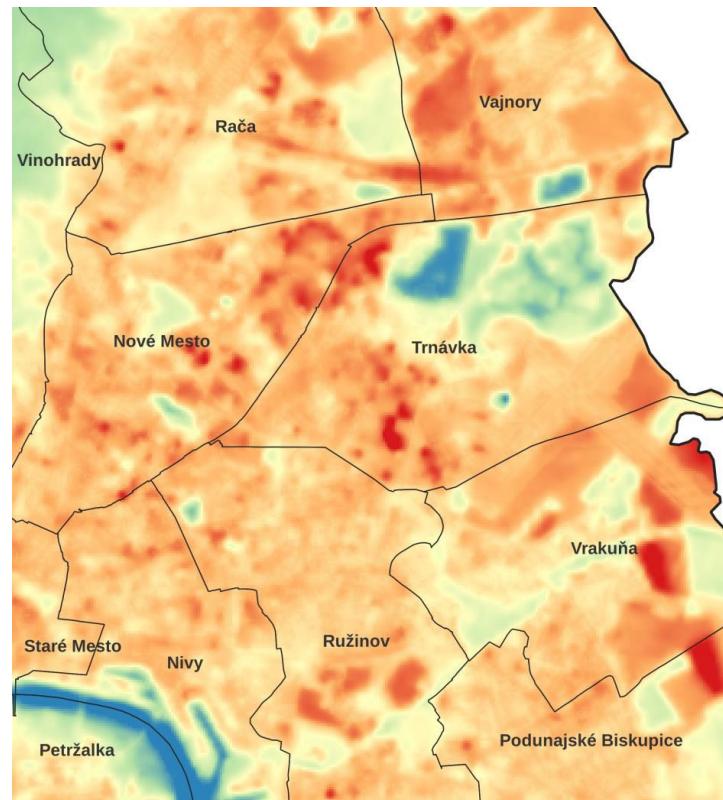
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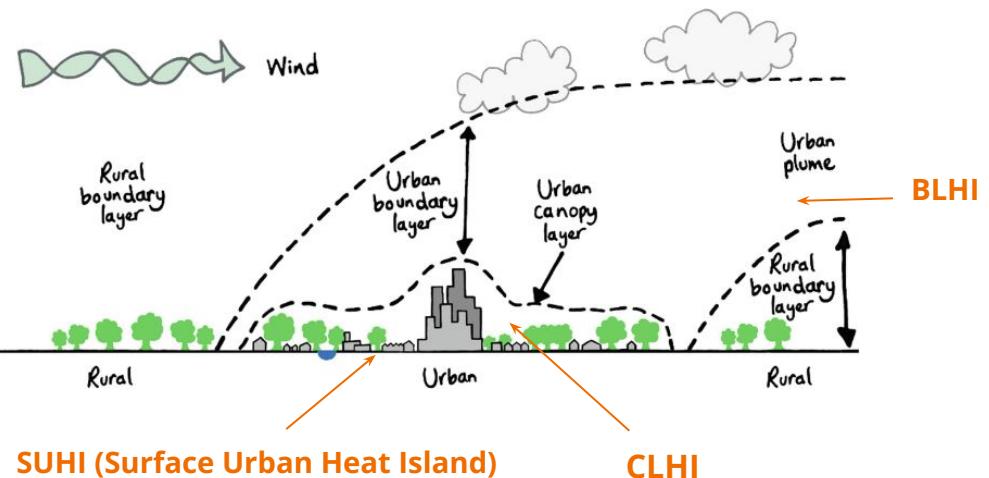
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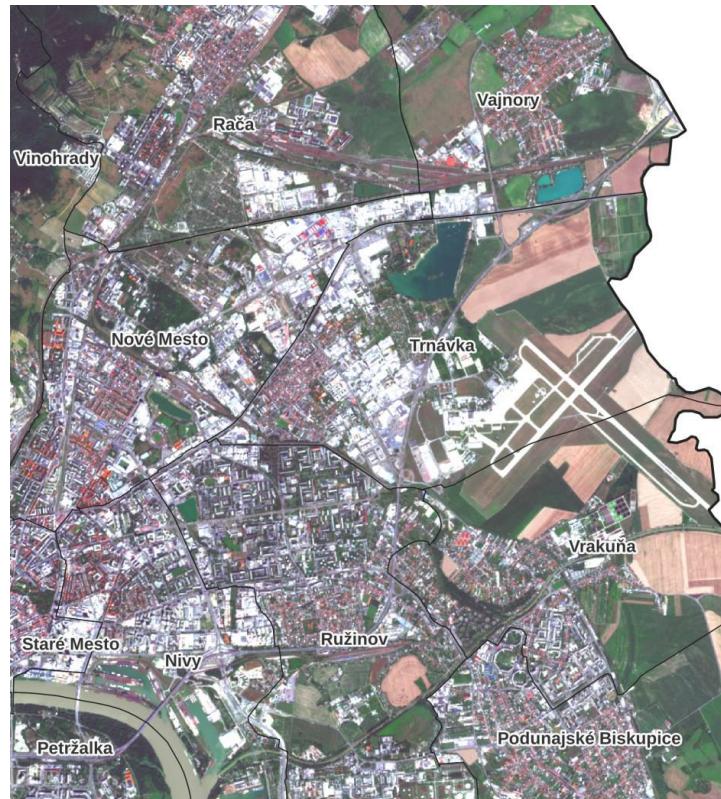
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Surface Temperature
in Bratislava
14.7.2022



Land Surface Temperature (LST)

Single-channel methods:

1. Radiative Transfer Equation (RTE) - Dash et al. (2001)

$$L_\lambda = \tau \varepsilon B_\lambda(T_s) + L_\lambda^\uparrow + \tau(1-\varepsilon)L_\lambda^\downarrow$$

Landsat Surface Temperature (ST) product

Landsat Collection 2 Level 2

2. Single-Channel Algorithm (SCA) - revised by Jiménez-Muñoz et al. (2009)

3. Mono-Window Algorithm (MWA) - Qin et al. (2001)

4. Statistical Mono-Window (SMW) - Sun et al. (2004)

$$LST = A_i \frac{Tb}{\varepsilon} + B_i \frac{1}{\varepsilon} + C_i$$

Google Earth Engine (GEE) code

Ermida et al. (2020)

5.

Split-window methods:

Temperature and Emissivity Separation (TES), Temperature Independent Spectral Index (TISI), ...

Land Surface Temper

Single-channel methods:

1. Radiative Transfer Equation (RTE)

$$L_\lambda = \tau \varepsilon B_\lambda(T_s) + L_\lambda^\uparrow + \tau(1-\varepsilon)L_\lambda^\downarrow$$

2. Single-Channel Algorithm (SCA) -

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$$LST = A_i \frac{Tb}{\varepsilon} + B_i \frac{1}{\varepsilon} + C_i$$

5.

Split-window methods:

Temperature and Emissivity Separatio

GeoKARTO 2024, September 5-6, 2024, Stará Lesná



Article

Google Earth Engine Open-Source Code for Land Surface Temperature Estimation from the Landsat Series

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Abstract: Land Surface Temperature (LST) is increasingly important for various studies assessing land surface conditions, e.g., studies of urban climate, evapotranspiration, and vegetation stress. The Landsat series of satellites have the potential to provide LST estimates at a high spatial resolution, which is particularly appropriate for local or small-scale studies. Numerous studies have proposed LST retrieval algorithms for the Landsat series, and some datasets are available online. However, those datasets generally require the users to be able to handle large volumes of data. Google Earth Engine (GEE) is an online platform created to allow remote sensing users to easily perform big data analyses without increasing the demand for local computing resources. However, high spatial resolution LST datasets are currently not available in GEE. Here we provide a code repository that allows computing LSTs from Landsat 4, 5, 7, and 8 within GEE. The code may be used freely by users for computing Landsat LST as part of any analysis within GEE.

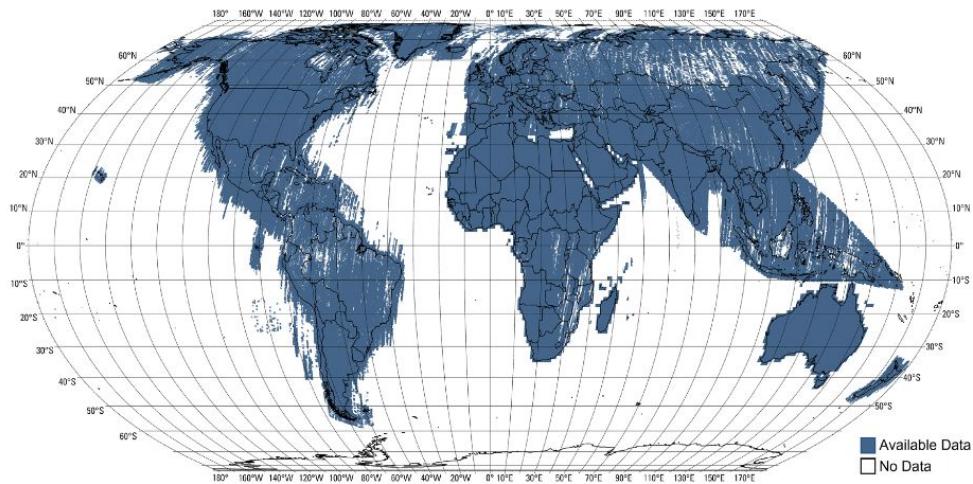
Land Surface Emissivity (LSE)

$$\varepsilon_\lambda(T) = \frac{L_\lambda(T)}{B_\lambda(T)}$$

ε - ratio of radiance emitted by a body at temperature T to the radiance of a black body at the same temperature

1. ASTER Global Emissivity Dataset (GED) v3

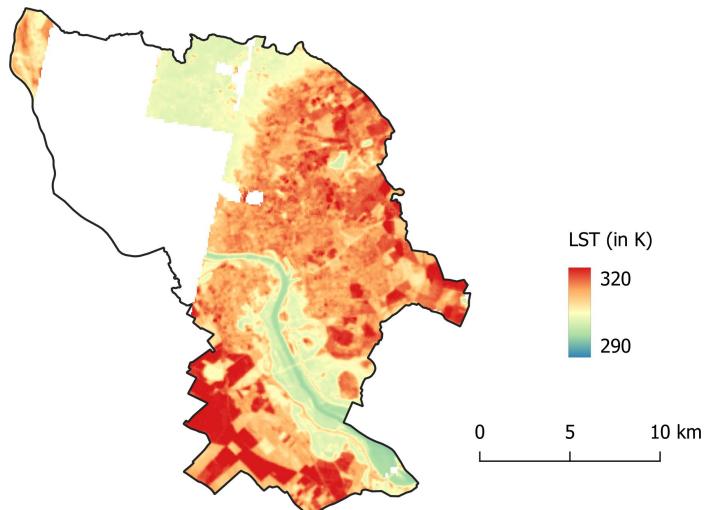
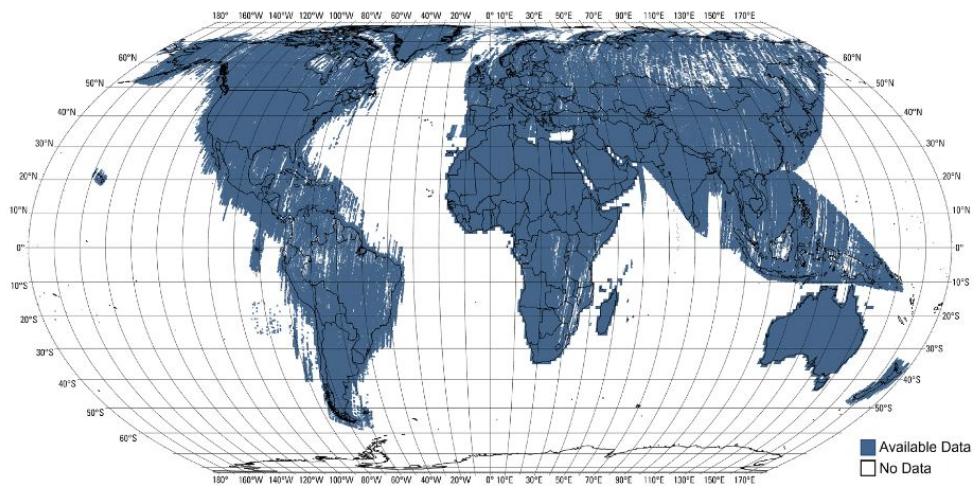
LST and LSE from ASTER 2000-2008 satellite scenes (all cloud-free pixels)



Land Surface Emissivity (LSE)

1. ASTER Global Emissivity Dataset (GED) v3

LST and LSE from ASTER 2000-2008 satellite scenes (all cloud-free pixels)



Land Surface Emissivity (LSE)

1. ASTER Global Emissivity Dataset (GED) v3

LST and LSE from ASTER 2000-2008 satellite scenes (all cloud-free pixels)

2. Classification-Based Emissivity Methods (CBEM) - Snyder et al. (1998)

3. NDVI-Based Emissivity Methods (NBEM) - Sobrino and Raissouni (2000), Sobrino et al. (2004)

NDVI-threshold method (NDVI^{THM})

$$\varepsilon = a_i \rho_{RED} + b_i$$

$$\varepsilon = \varepsilon_v FVC + \varepsilon_s (1 - FVC) + C_i$$

$$\varepsilon = \varepsilon_v + C_i$$

Simplified NDVI-threshold method (SNDVI^{THM})

$$\varepsilon = \varepsilon_s$$

$$\varepsilon = \varepsilon_s FVC + (\varepsilon_v - \varepsilon_s) FVC$$

$$\varepsilon = \varepsilon_v$$

$$NDVI < NDVI_s$$

$$NDVI_s \leq NDVI \geq NDVI_v$$

$$NDVI > NDVI_v$$

Bare soil

Mixed

Vegetation

$\varepsilon_s, \varepsilon_v$ - emissivity of soil and vegetation

$NDVI_s, NDVI_v$ - threshold values for soil and vegetation

Land Surface Emissivity (LSE)

1. ASTER Global Emissivity Dataset (GED) v3

LST and LSE from ASTER 2000-2008 satellite scenes (all cloud-free pixels)

2. Classification-Based Emissivity Methods (CBEM)

RTE method, SMW method

3. NDVI-Based Emissivity Methods (NBEM)

Method	Reference	Spectral library
NDVI ^{THM} SO	Sobrino et al. (2008)	ASTER
NDVI ^{THM} SK	Skoković et al. (2014)	ASTER
NDVI ^{THM} YU	Yu et al. (2014)	MODIS
SNDVI ^{THM} SK	$\varepsilon_s, \varepsilon_v$ from Skoković et al. (2014)	ASTER
SNDVI ^{THM} YU	$\varepsilon_s, \varepsilon_v$ from Yu et al. (2014)	MODIS
SNDVI ^{THM} WA	$\varepsilon_s, \varepsilon_v$ from Wang et al. (2015)	ASTER

GEE code

GEE Code Editor

Google Earth Engine Run

Scripts Docs Assets

LST_RTE_NDVI_add *

```
82 |
83 // calculate emissivity
84 var landsatSR = ee.ImageCollection(scenes.map(function(id) {
85   return ee.Image(coll_SR+ '/' +id);
86 }))
87 .select(SR_bands)
88 .map(func.calNDVI)
89 .map(func.calcFVC)
90 .map(func.calcCI)
91 .map(func.calcESV)
92 .map(func.calLSE) //SNDVithm method
93 .map(func.calLSE2) //NDVithm method
94
95 // create TIR image collection from SR collection
96 var filterDNBySceneIDs = function(image) {
97   var sceneID = image.get('system:index');
98   return ee.ImageCollection(coll_DN).filterMetadata('system:index', 'equals', sceneID).select
99 };
100
101 var filteredDN = landsatSR.map(filterDNBySceneIDs);
102
103 // calculate TOA
104 var landsatTOA = filteredDN
105 .map(func.calCLTOA)
106 print('LandsatTOA', landsatTOA)
```

Inspector Console Tasks

Use `print(...)` to write to this console.

LandsatTOA
↳ ImageCollection (1 element)

774.8853
1321.0789

landsatSR
↳ ImageCollection (1 element)

landsatLST
↳ ImageCollection (1 element)

landsatTOA
↳ ImageCollection (1 element)

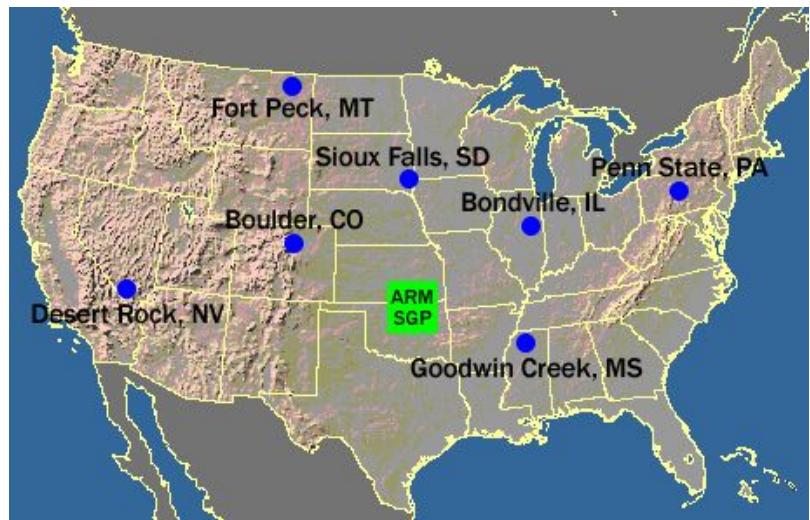
landsatLST
↳ ImageCollection (1 element)

LST validation



SURFRAD (Surface Radiation Budget) Network

- broadband downwelling and upwelling IR radiance

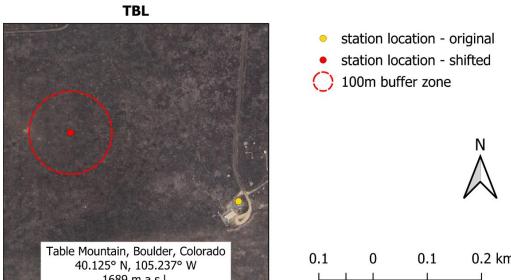
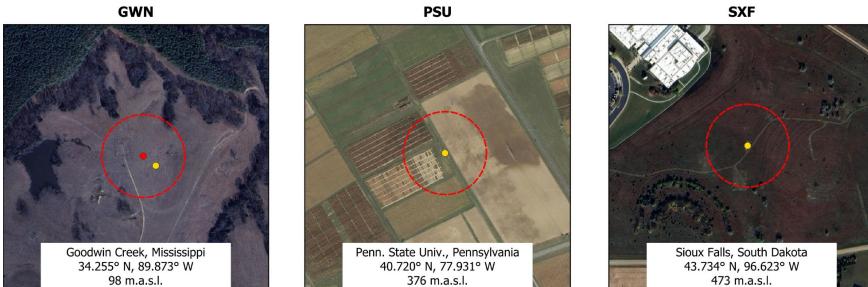
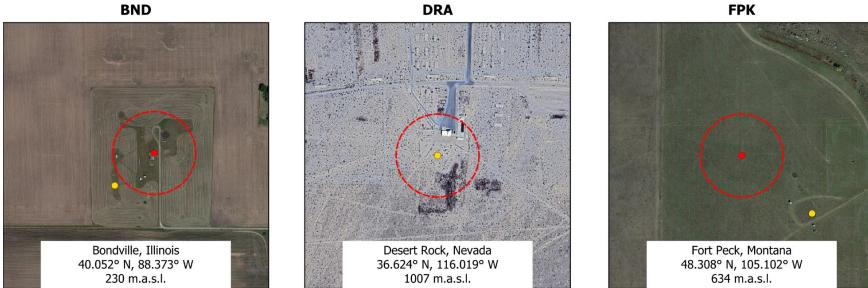
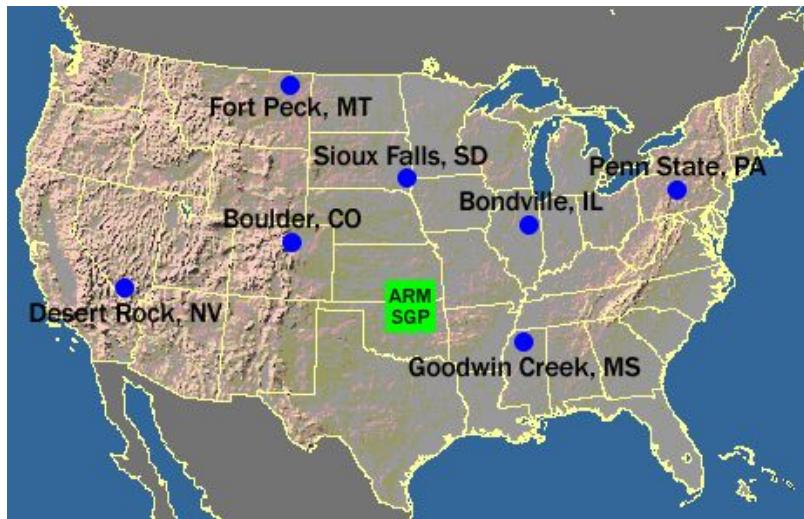


LST validation



SURFRAD (Surface Radiation Budget) Network

- broadband downwelling and upwelling IR radiance



Results - Landsat 8

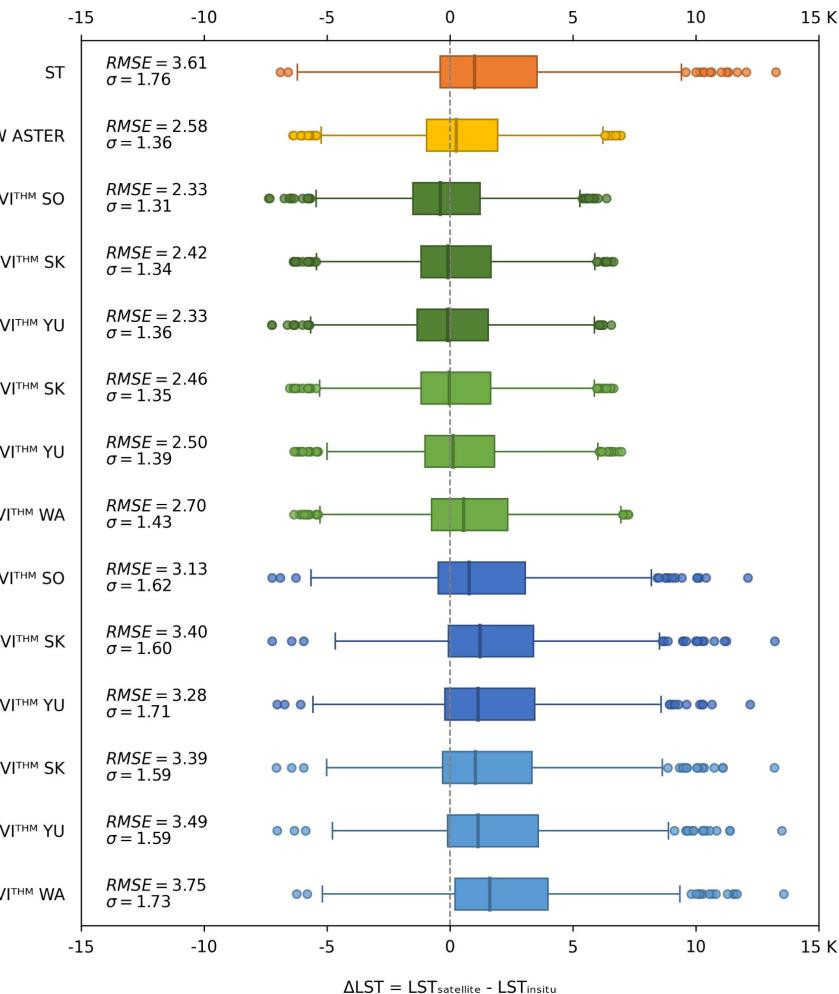
All Landsat 8 scenes with cloud cover < 11%

Validation statistics:

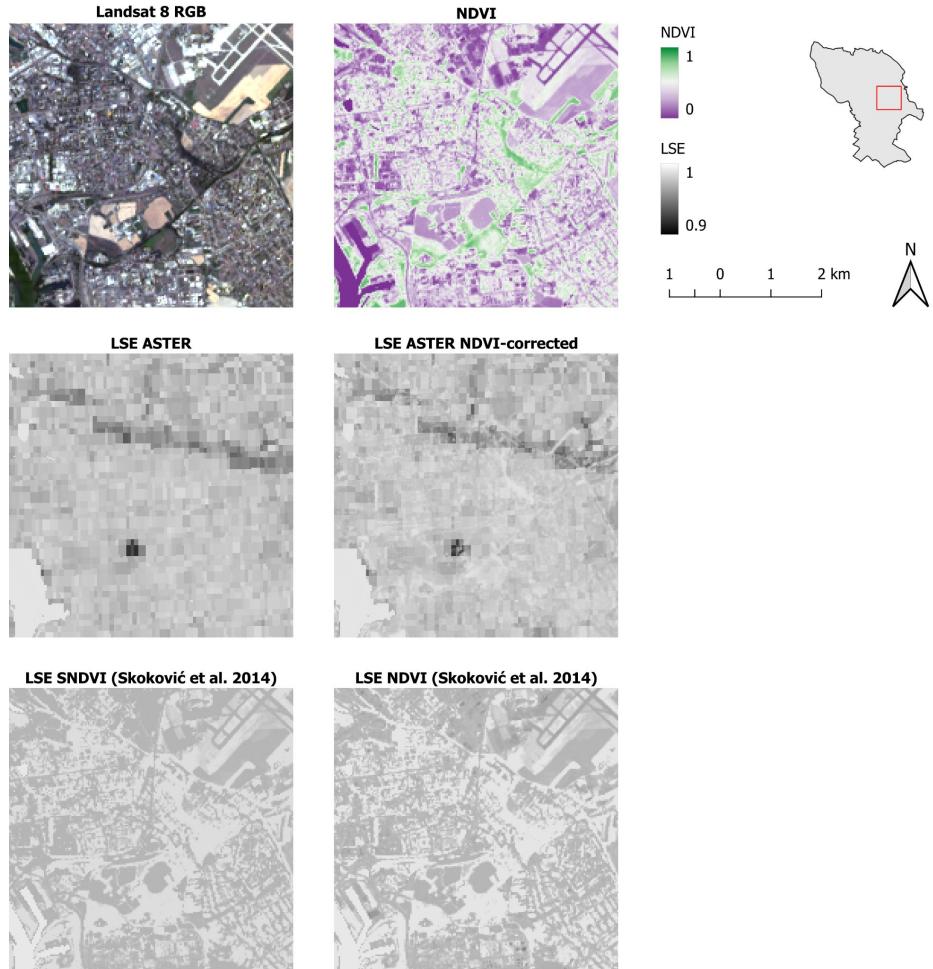
RMSE, accuracy/bias μ , precision σ

3 σ -Hampel identifier to exclude outliers

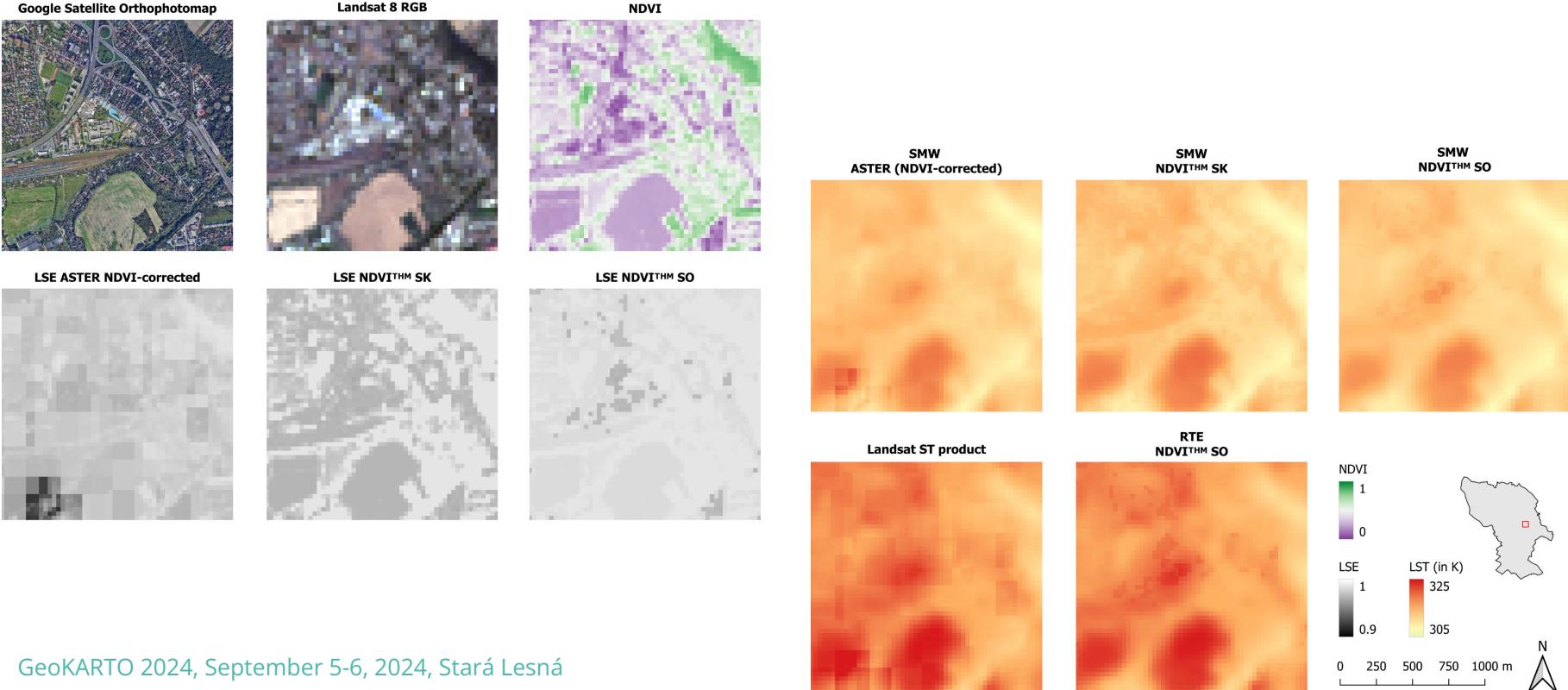
N	ST ASTER	SMW ASTER	LSE	RMSE (in K)						
				NDVI ^{THM}			SNDVI ^{THM}			
				SO	SK	YU	SK	YU	WA	
882	3.61	2.58	SMW	2.33	2.42	2.33	2.46	2.50	2.70	
951	7.15	6.18		6.08	6.10	6.05	6.16	6.16	6.22	
RTE				3.13	3.40	3.28	3.39	3.49	3.75	
				6.90	7.00	6.92	7.04	7.07	7.18	



Results - LSE

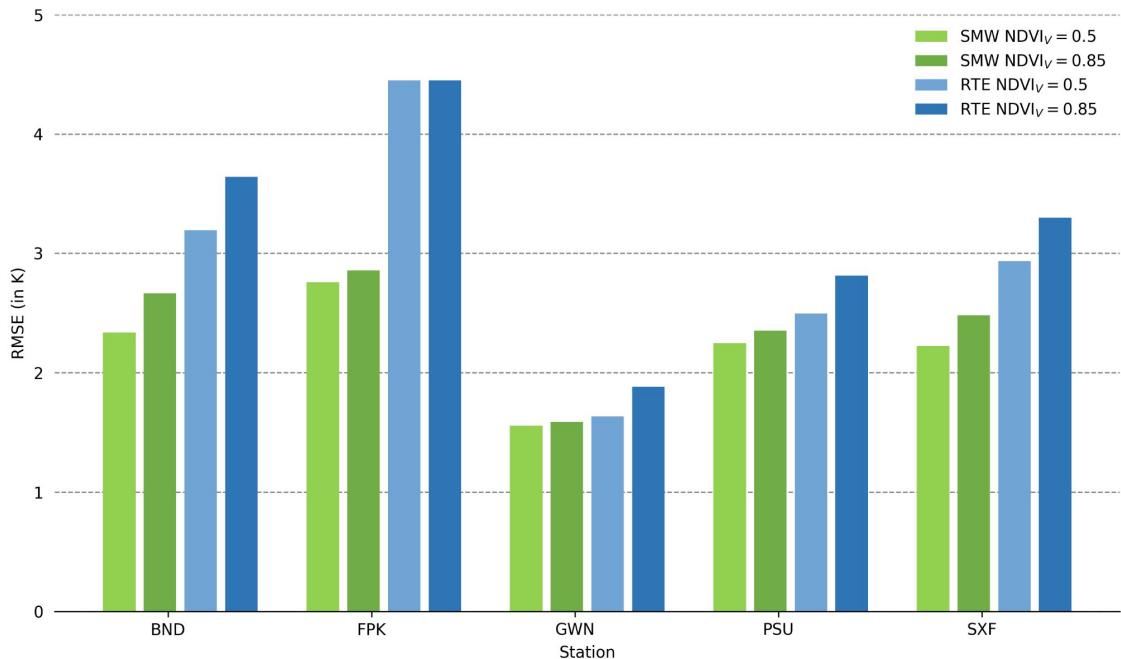


Results - LSE



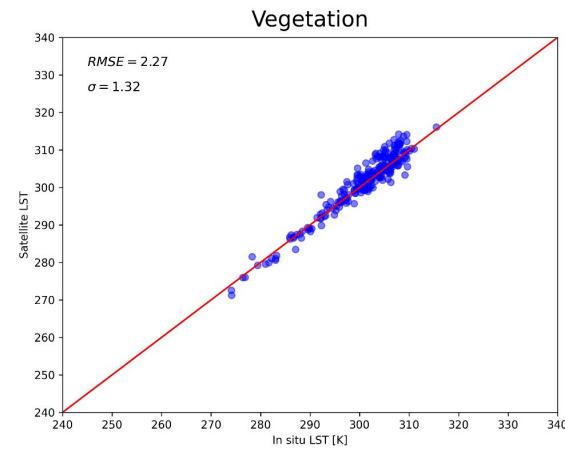
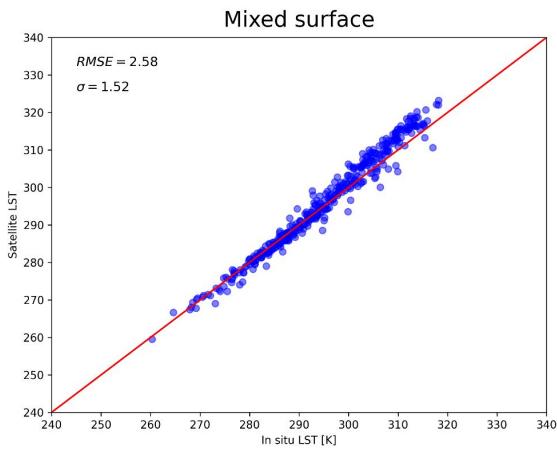
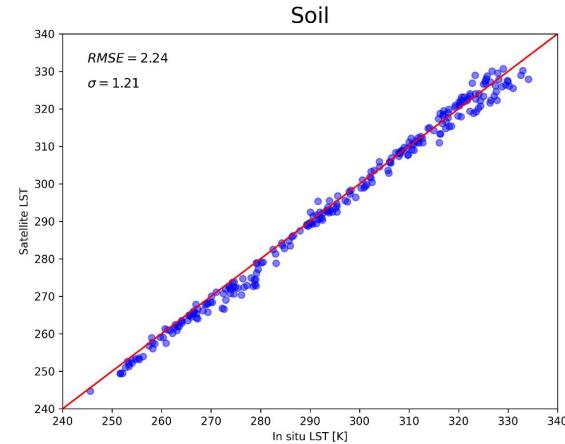
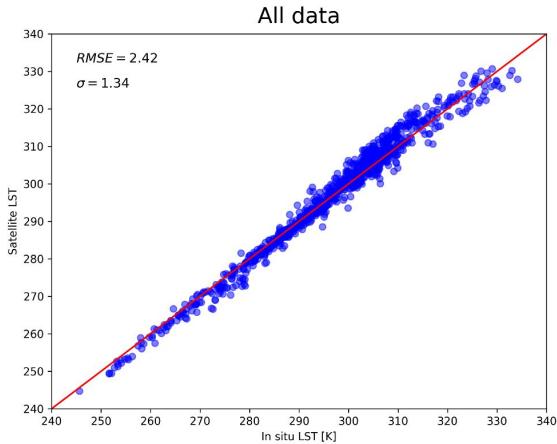
Results - NDVI

$NDVI_v = 0.5$ (Sobrino and Raissouni 2000; Sobrino et al. 2008; Yu et al. 2014;
Ndossi and Avdan 2016; Sekertekin and Bonafoni 2020)
 $NDVI_v = 0.85$ (Jiménez-Muñoz et al. 2009; Parastatidis et al. 2017; Ren et al. 2017;
Ermida et al. 2020)



Results - NDVI

Soil	$\text{NDVI} < 0.2$
Mixed surface	$0.2 \leq \text{NDVI} \leq 0.5$
Vegetation	$\text{NDVI} > 0.5$



Results - Landsat 5, Landsat 7, Landsat 9

RMSE (in K)

Mission	N	ST	SMW	SMW		RTE	
				NDVI ^{THM}	SNDVI ^{THM}	NDVI ^{THM}	SNDVI ^{THM}
Landsat 5	1100	3.47	2.67	2.36	2.37	3.20	3.19
	1148	4.56	3.88	3.66	3.70	4.31	4.35
Landsat 7	1563	3.58	2.68	2.44	2.49	3.20	3.20
	1666	6.59	5.83	5.69	5.77	6.35	6.43
Landsat 8	882	3.61	2.58	2.42	2.46	3.40	3.39
	951	7.15	6.18	6.10	6.16	7.00	7.04
Landsat 9	202	3.92	2.78	2.62	2.65	3.85	3.83
	212	5.33	4.33	4.23	4.27	5.28	5.29

Conclusions

- Operational LST calculation method from Landsat 4-5, 7, 8 and 9, implemented in GEE code repository
- Filling gaps and avoiding block artefacts in ASTER GED v3
- Considering the current state of the vegetation at the time of imaging
- Higher accuracy and precision than original SMW method and the USGS Landsat ST product
- More accurate mapping of UHI on a local and regional scale

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Thank you for your attention !

hana.bobalova@uniba.sk