Assessing Water Ecosystem Services and their Dependency on Land use changes A Case Study of the Azov sea basin

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Project

Part of Environmental Systems laboratory at CEU,

Budapest, Hungary

Ongoing since 2012, work in progress

With Viktor Lagutov & Irina Gilfanova

Focussed research on the Azov sea Basin



Project – Environmental Systems Laboratory at CEU

Project aim:

To assess the development of Ecosystem Goods & Services

in the Azov Sea Basin with ArcSWAT Modelling

• My Focus:

To analyse Land Use and Land Cover changes in the basin

with Google Earth Engine as an input to the project

The Azov Sea Basin



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(Lagutov and Lagutov 2010)



(Keeler et al. 2012)

"What has been the influence of land use and land cover changes on ecosystem goods and services of the Azov Sea basin in the past 30 years, and how could this relationship develop in the future?"

Methodology Google Earth Engine



Methodology Google Earth Engine

Supervised Landuse classification

• Using Landsat 8 images of August of different years,

clipped to the borders of the basin

andus	semap2015_landsat8	Get Link	Save 👻	Run	Reset 👻	\$
1	// Look for adequate image filtered by date and cloud cover	and regi	on			*
2	<pre>var Azov_sea = ee.ImageCollection('LANDSAT/LC08/C01/T1_RT_T</pre>	'0A')				
3	.filterBounds(shape_basin)					
4	.filterDate('2015-08-01','2015-09-01');					
5						
6	<pre>var sorted = Azov_sea.sort('CLOUD_COVER', false).mosaic();</pre>					
7						
8	<pre>print(sorted);</pre>					
9						
10	//clip it to the exact region					
11	<pre>var clipped = sorted.clip(shape_basin);</pre>					-
10						*

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Creating Training Data

Least cloudy picture available for time frame for every single point chosen

Band combination 6-5-4 used for vegetation analysis

Landsat 8	Bands	Wavelength (micrometers)
Operational		(merometers)
Land Imager	Band 1 - Ultra Blue (coastal/aerosol)	0.435 - 0.451
(OLI)	Band 2 - Blue	0.452 - 0.512
and	Band 3 - Green	0.533 - 0.590
Thermal	Band 4 - Red	0.636 - 0.673
Infrared	Band 5 - Near Infrared (NIR)	0.851 - 0.879
Sensor	Band 6 - Shortwave Infrared (SWIR) 1	1.566 - 1.651
(TIRS)	Band 7 - Shortwave Infrared (SWIR) 2	2.107 - 2.294
	Band 8 - Panchromatic	0.503 - 0.676
	Band 9 - Cirrus	1.363 - 1.384
	Band 10 - Thermal Infrared (TIRS) 1	10.60 - 11.19
	Band 11 - Thermal Infrared (TIRS) 2	11.50 - 12.51



(Barsi et al. 2014)

Creating Polygons



Training classifier & Image classification

```
41 // Create a random forest classifier with custom parameters.
42 var classifier = ee.Classifier.randomForest().train({
      features: training,
43
                                                                          Different classifier types
      classProperty: 'class',
44
      inputProperties: bands
45
    });
46
47
    // Train the classifier.
48
    var trained = classifier.train(training, 'class', bands);
49
50
   // Classify the image.
51
    var classified = clipped.classify(trained);
52
53
   // Create a palette to display the classes.
54
    var palette =['0000FF', 'ffa500','000000','FFFF00','00FF00','f4f4f4'];
55
56
57 //0000FF is water is blue
  //ffa500 is cropland with vegetation is orange
58
59 //00000 is urban is black
60 //ffff00 is bare land is yellow
61 //00ff00 is natural vegetation is green
62
   // Display the classification result and the input image.
63
    Map.addLayer(classified, {min: 0, max: 5, palette: palette}, 'Landuse Type 2015');
64
CE.
```

Preliminary Results

Land use analysis Google Earth Engine



Blue:

Urbanised areas Black: Naturally vegetated areas Green: Cropland Orange: Yellow: Pasture/livestock Water



Land use analysis Google Earth Engine

ОгуоІ Орёл

Poltava Полтава

Zaporizhzhia

Novoros

Новоросо

уууі Rih



Black: Urbanised areas Naturally vegetated areas Green: Orange: Cropland Yellow: Pasture/livestock Blue: Water





Accuracy Assessment

Legend:

0 = Water

- 1 = Vegetated Cropland
- 2 = Urban land
- 3 = Bare cropland/Pasture
- 4 = Natural vegetation/forest

Inspector Console Tasks	
Resubstitution error matrix:	150N
Thist (F alamants)	3500
	120M
▶0: [139949,4,4,0,6]	
▶ 1: [9,9279,91,37,122]	
▶ 2: [11,134,2753,75,39]	
▶ 3: [2,65,31,12232,25]	
▶ 4: [6,113,55,27,11834]	
Training overall accuracy:	JSON
0.9951611900306948	



Issues to solve

Cloudless pictures vs. same month of the year

More detailed land use training data for different years

Separate accuracy assessment data

Creating an input for ArcSWAT

References

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Thank you for your attention!