Comparison of various types of land use and land cover data and example of their harmonisation

Lena Halounová, ISPRS Secretary General, CTU in Prague



Land Use and Land Cover

Land use corresponds to the socio-economic description i.e., functional dimension of areas:

areas used for residential, industrial or commercial purposes, for farming or forestry, for recreational or conservation purposes, etc.







Land Use and Land Cover



the arrangements, activities and inputs people undertake in a certain land cover type to produce, change or maintain it.

- "grassland" is a cover term,
- "rangeland" or "tennis court" refer to the use of a grass cover;

Food and Agriculture Organization of the United Nations (FAO)

Land Use and Land Cover

Food and Agriculture Organization of the United Nations (FAO)



Land cover is the observed (bio)physical cover on the earth's surface.

it should be confined to the description of vegetation and man-made features.

Also, it is disputable whether **water surfaces** are real land cover.

However, in practice, the scientific community usually includes these features within the term *land cover*.

Land Use and Land Cover

Links with land cover are possible; it may be possible to infer land use from land cover and conversely.

But situations are often complicated and the link is not so evident.

Contrary to land cover, land use is difficult to 'observe'.

European Environment Agency



Land Use or Land Cover?

Do we separate them in the processed data?

Yes, if we use appropriate nomenclature

To be able to decide we have to have a look at existing land use/ land cover data

Geoharmonizer: EU-wide automated mapping system for harmonization of Open Data based on FOSS4G and Machine Learning



https://opendatascience.eu/geoharmonizer-project/



Geoharmonizer: EU-wide automated mapping system for harmonization of Open Data based on FOSS4G and Machine Learning

overall objective of the Action

data portal and a software suite

free and open-source software for geospatial data (FOSS4G)using **Machine Learning** algorithms

coverage of the entire Europe

all input data are free data

Geoharmonizer: EU-wide automated mapping system for harmonization of Open Data based on FOSS4G and Machine Learning

European Land Cover from satellite data

one of the main outcomes

2000 - 2020

Land Use /Land Cover data were regarded from various views

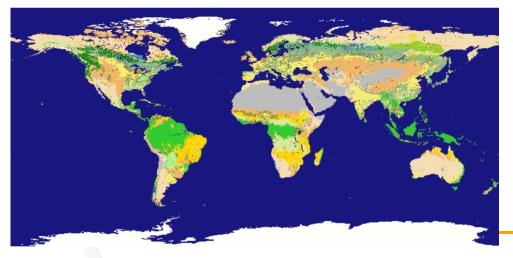
spatial coverage thematic resolution spatial resolution temporal coverage

Spatial coverage

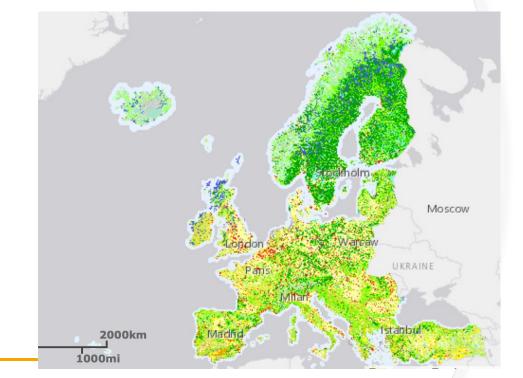
Global, Continental, National and Regional Land Cover products



LAND COVER CLASSIFICATION (1 YEAR)



NASA Land Cover (2011)



CORINE Land Cover 2018, Copernicus

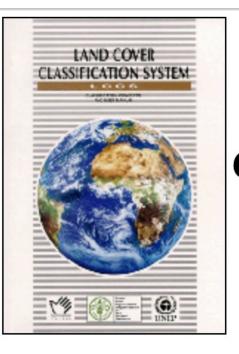
Spatial Coverage

Global data

LAND COVER NAME	MAINTAINER
CCI Land Cover (2017)	ESA
<u>GlobeLand30 (</u> 2000, 2010)	National Geomatics Center of China
<u>GLC - SHARE</u>	Land and Water Division of FAO
<u>GLC2000 (2000)</u>	Joint Research Center (JRC)
NASA Land Cover (2011)	NASA
Land Cover (GLCNMO)	ISCGM, GSI of Japan
<u>Copernicus Global Land</u> <u>Service</u>	VITO

Global data

Product / Characteri stics	ESA CCI LC	GlobeLand3 0	<u>GLC -</u> <u>SHARE</u>	<u>GLC2000</u>	NASA LC	<u>Land Cover</u> (<u>GLCNMO)</u>	<u>CGLOPS-</u> <u>1 (VITO)</u>
Thematic resolution (No. of classes)	22	10	11	23	17	20	23
Nomencla ture model	LCCS	GlobeLand3 0	LCCS, FAO, SEEA	LCCS	<u>IGBP</u>	LCCS	LCCS
Spatial resolution / MMU	300m	30 m	1 km (<u>source</u>)	1 km (<u>source</u>) (<u>source</u>)	1 deg 0.5 deg 0.25 deg	500 m	100 m



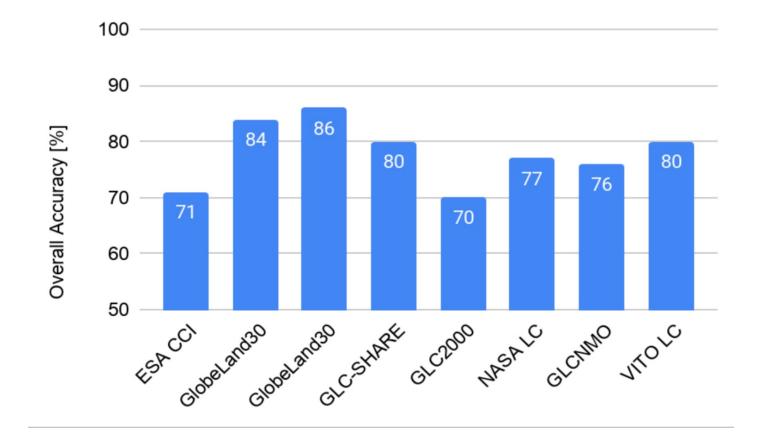


Nomenclature:

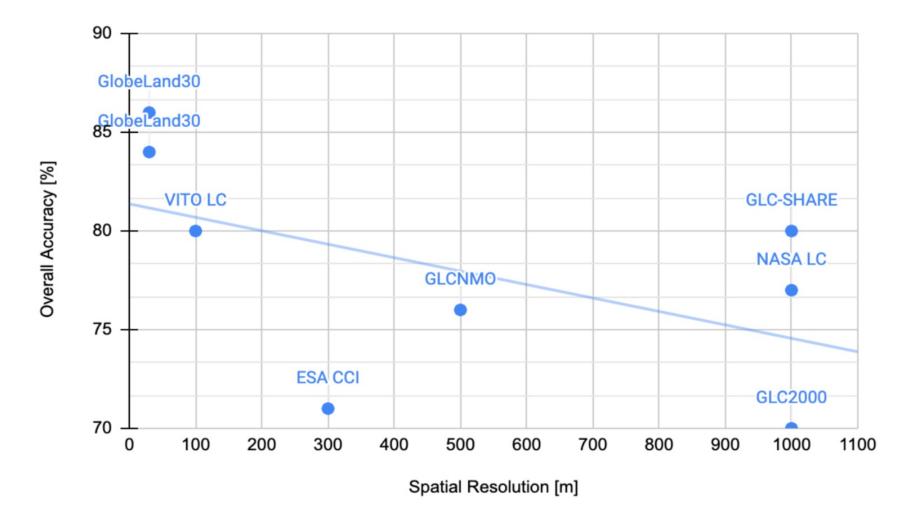
- LCCS (FAO)http://www.fao.org/3/x0596e/x0596e00.htmGlobeLand30 http://www.globeland30.org
- IGBP (International Geosphere-Biosphere Programme) http://www.igbp.net



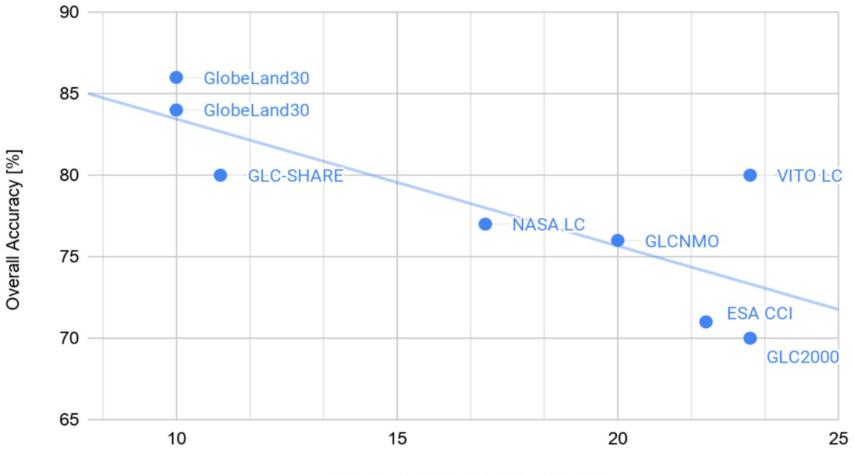
Global data - overal accuracy



Global data - spatial resolution vs. overal accuracy

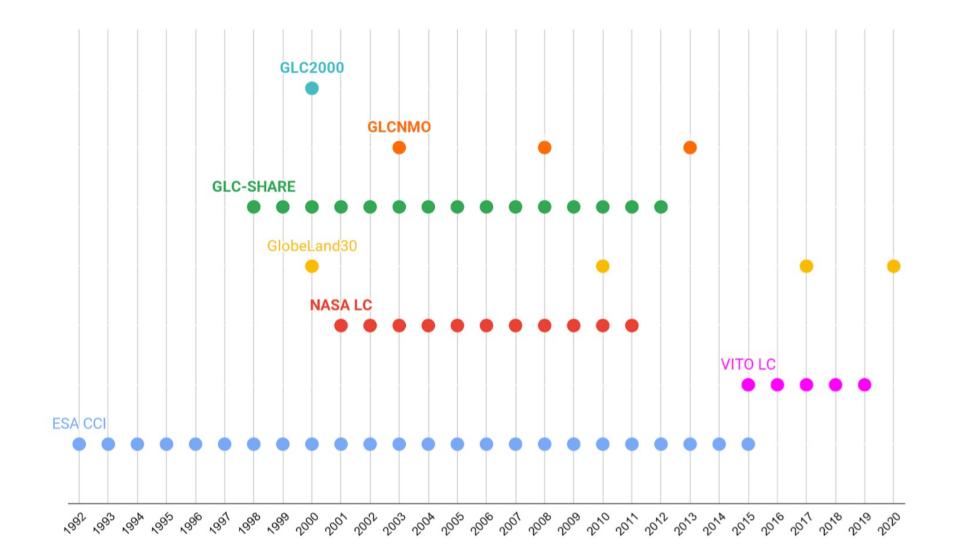


Global data - thematic resolution vs. overal accuracy



Thematic Resolution [No. classes]

Global data – temporal coverage



European – continental data

Land Cover name	Maintainer
CORINE Land Cover (1990 - 2018)	EEA (EU Copernicus)
European Settlement Map (2016, 17) Urban Atlas	EEA (EU Copernicus)
Land Cover Map of Europe 2017 (S2GLC, ESA)	EEA (EU Copernicus)
Pan-European land cover (30 m) map of 2015 (based on Landsat and LUCAS data)	Pflugmacher, Rabe, Peters, Hostert (Humbold University, Berlin)
High Resolution Layers	Copernicus

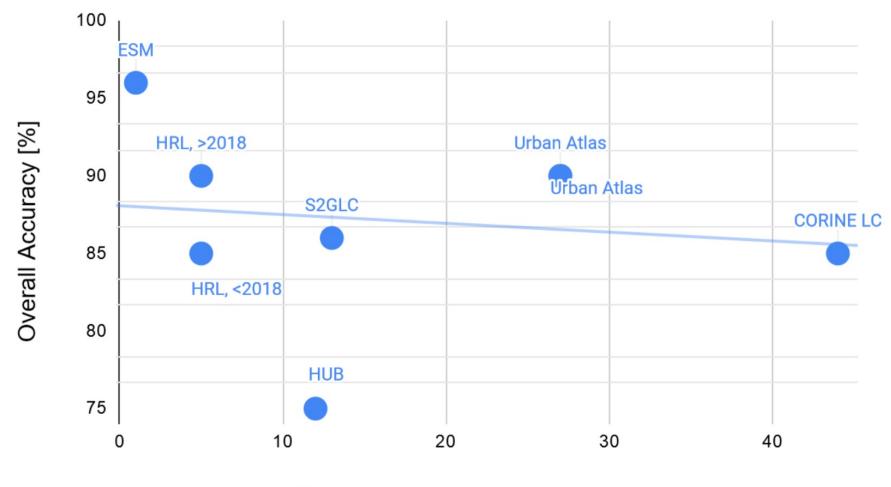
European data LC/LU

Product\ Characteristic s	CORINE Land Cover (Copernicus)	<u>Global Land</u> <u>Cover -</u> <u>Sentinel- 2</u> <u>S2GLC (ESA)</u>	<u>Urban Atlas</u> (Copernicus)	European Settlement <u>Map</u> (Copernicus)	Pan-European LC (Humboldt University of Berlin)	<u>High</u> <u>Resolution</u> <u>Layers (HRL)</u> (Copernicus)
Spatial coverage	Europe (EEA39)	Europe	Europe (EEA39) 319 FUAs	Europe	Europe	Pan-Europea n
Thematic resolution (No. of classes)	44	13	27 (17 urban classes 10 rural classes)	1	12	5
Nomenclature model	CORINE (LC/LU)	CORINE	CORINE compatible	-	CORINE compatible	CORINE compatible
Spatial resolution (pixel size)		10 m		2 m/10 m/100 m	30 m	10 m (20 m before 2018)

European data - overal accuracy

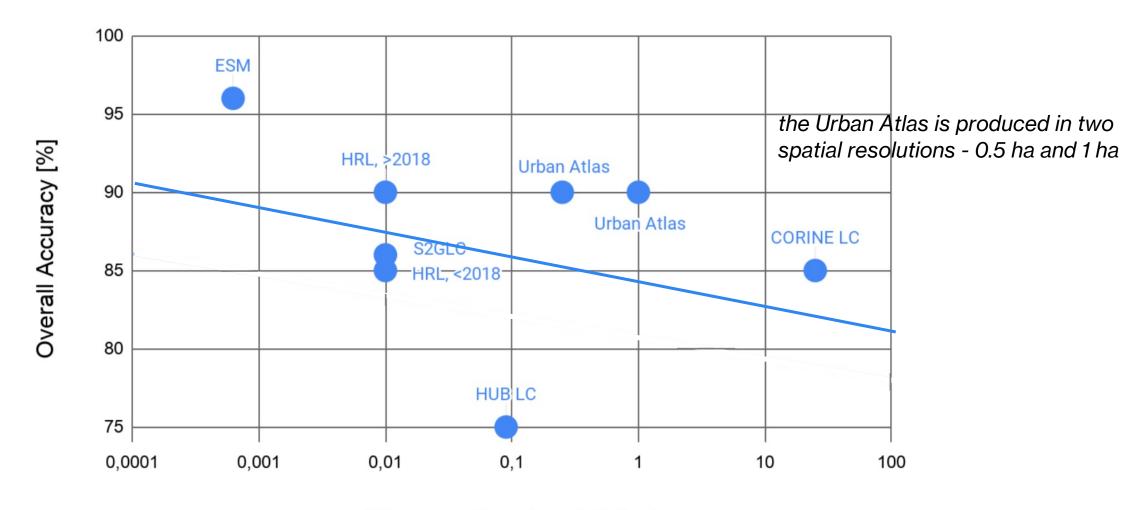


European data thematic resolution vs. overal accuracy



Thematic Resolution [№ of classes]

European LC data spatial resolution and overal accuracy



Minimum Mapping Unit [ha]

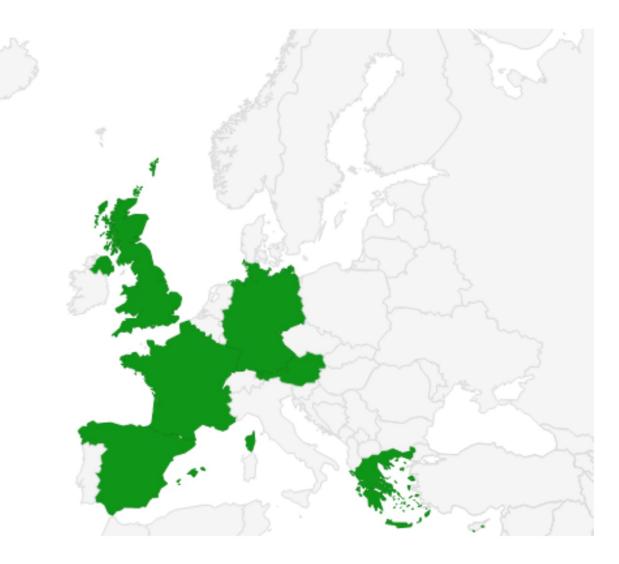
European data – temporal coverage



National Land Cover data – publicly available

Land Cover name	Maintainer
LISA	Austria, AT consortium
LandCover DE	Germany, DLR
SIOSE	Spain, National Reference Center on Land Cover and on Land Use and Spatial Planning
LGN (Dutch land use datasets from 1995 to 2018)	Netherlands, Wageningen Environmental Research
OSO Land Cover	France, Theia Data and Services centre
LC Greece	Greece, University of the Aegean
<u>CLC50</u>	Hungary, Institute of Geodesy, Cartography and Remote Sensing (FÖMI),

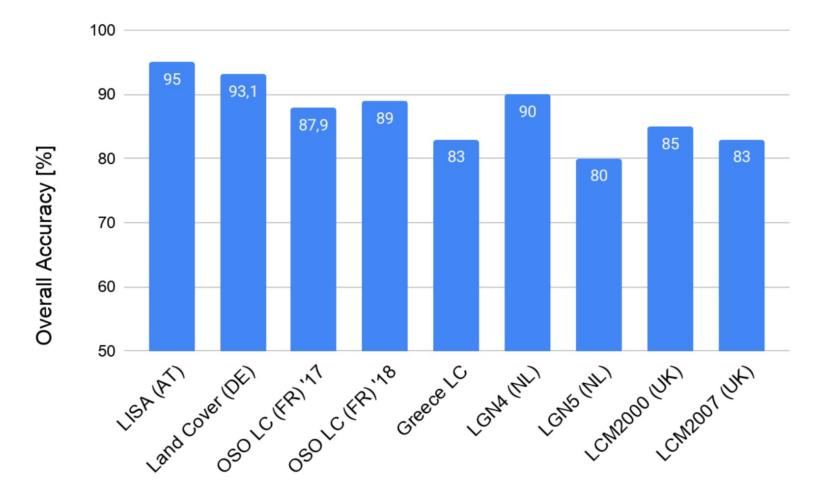
Map of publicly available European national data



National Land Cover data – publicly available

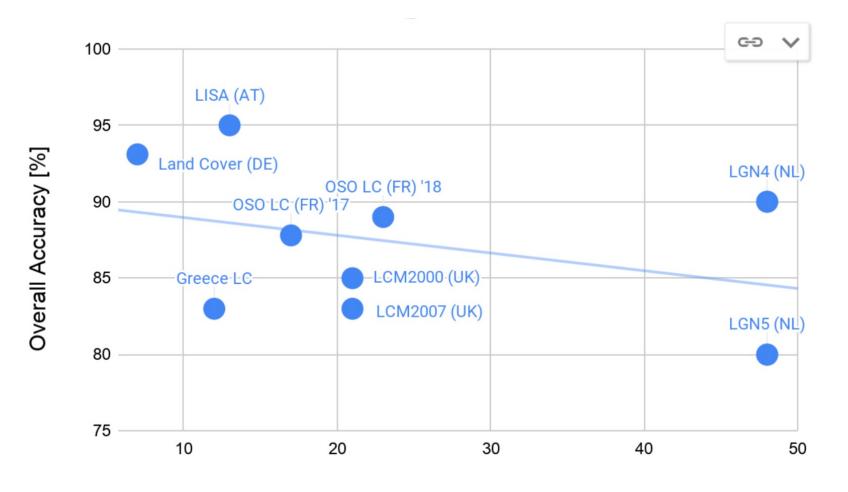
Product / Charact eristics	<u>LISA</u>	Land Cover DE	<u>SIOSE</u>	<u>OSO LC</u>	LC Greece	<u>CLC50</u> <u>HU</u>	<u>LGN</u>	LCM2015
Spatial coverag e	Austria	German y	Spain	France	Greece	Hungary	The Netherland s	UK
Themati c resoluti on (No. of classes)	13 and 12 attribute s	7	20	23 (17 before 2018)	12	79	48 (39 before 2018)	21
Nomen clature model	CORINE compati ble	CORINE	SIOSE	OSO LC	CORINE	CORINE level-3	LGN	JNCC Broad Habitats

National European data - overal accuracy



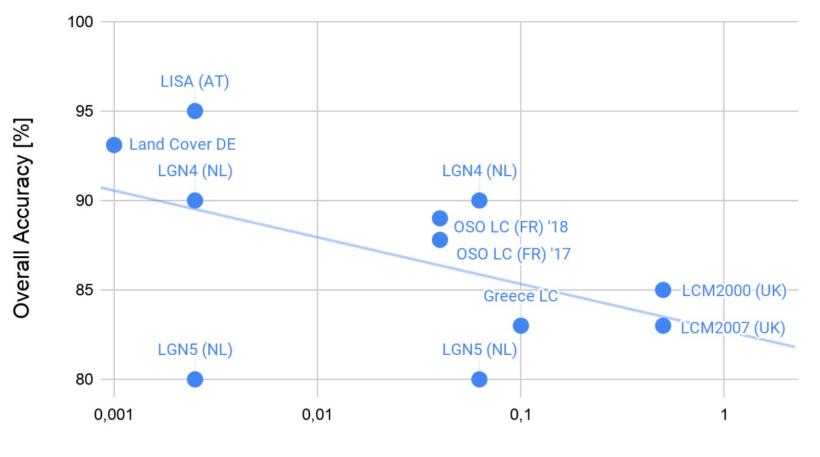
SIOSE products do not report overall accuracy

National European data thematic resolution vs. overal accuracy



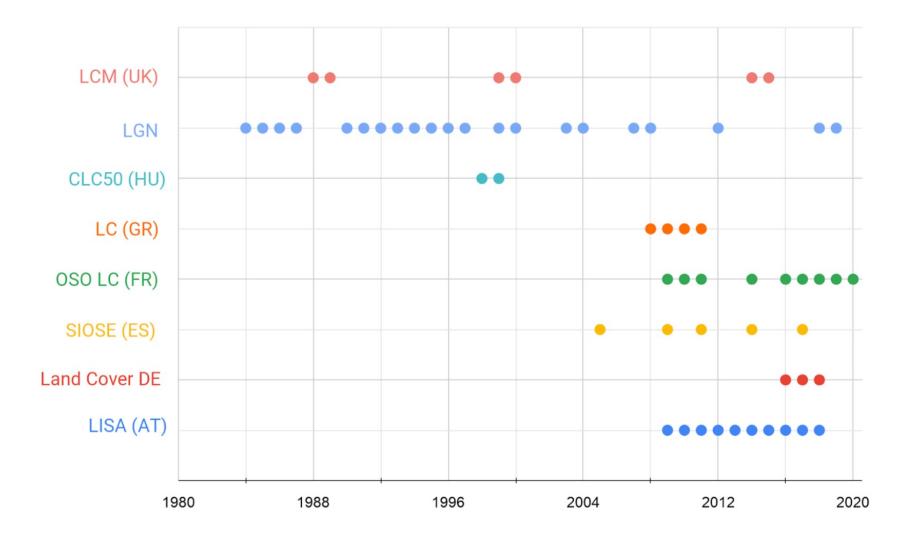
Thematic Resolution [№ of classes]

National European data spatial resolution vs. overal accuracy

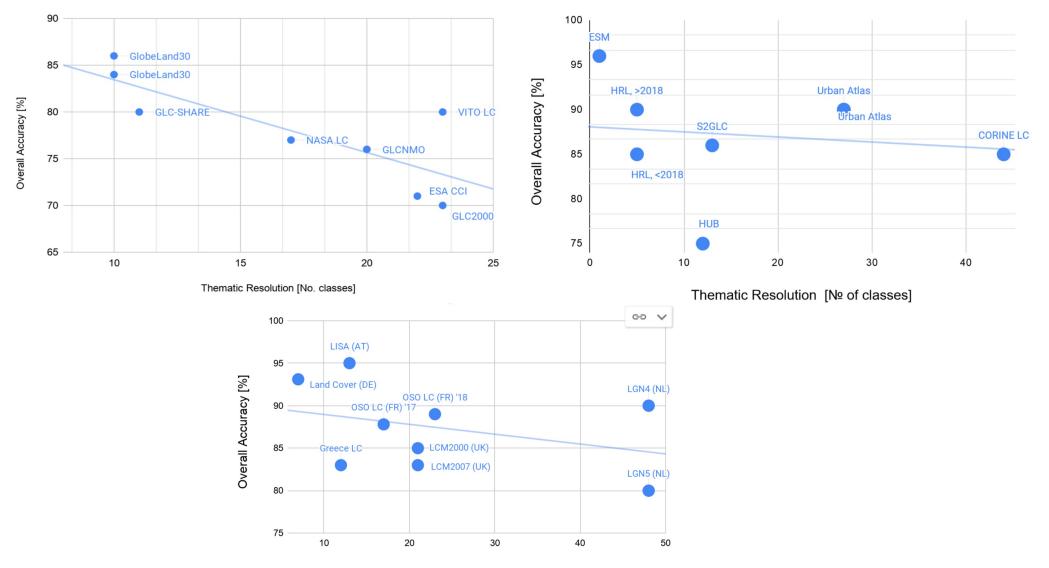


Spatial Resolution [ha]

National European data – temporal coverage



Comparison thematic resolution and overal accuracy for global, continental and national European data



Thematic Resolution [№ of classes]

Conditions for good results of (land cover) classification

Input data

- remote sensing data wide choice
 - good quality
 - representative spectral information
 - appropriate spatial resolution
 - necessary temporal resolution
- all other spatial data which are available

Method incl. sample /training data

Software

Data used for the spatio-temporal data land cover/land use European model of Geoharmonizer

Land Cover DataCORINEImage: Constraint of the second seco

Satellite Data - open data

Landsat data

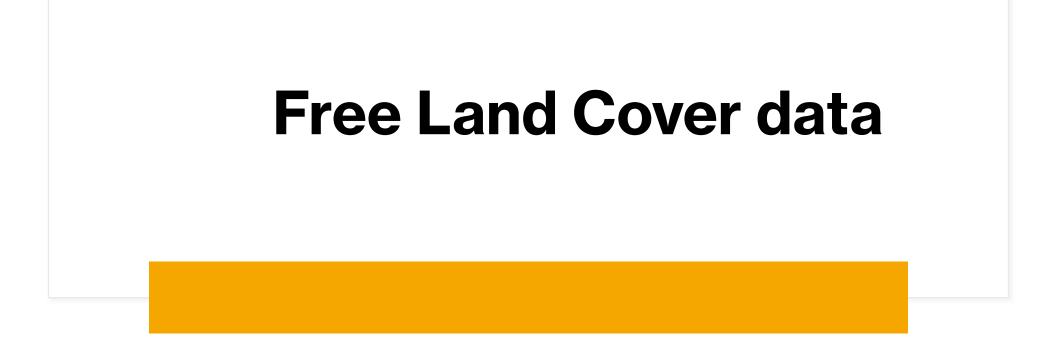
Sentinel 2

VIIRS/Suomi NPP night light Other data - timeless

DTM elevation DTM slope

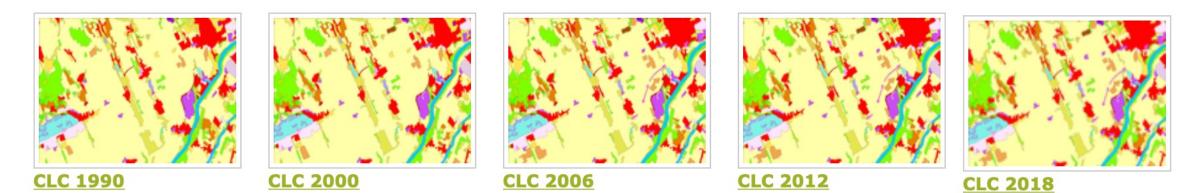
Monthly Geometric Temperature

Global Surface Water Historical occurrence (1984-2019)





Year	Input data	Input Data Resolution	Participating Countries
CLC1990	Landsat-5 MSS/TM	50 m	27
CLC2000	Landsat-7 ETM	25 m	30
CLC2006	SPOT - 4/5 and ORS P6 LISS	25 m	38
CLC2012	IRS P6 LISS II and RapidEye	25 m	
CLC2018	Sentinel-2 imagery data; Landsat-8	10 m	39



https://land.copernicus.eu/pan-european/corine-land-cover



1. Artificial surfaces

1.1 Urban fabric

1.1.1. Continuous urban fabric

1.1.2. Discontinuous urban fabric

1.2 Industrial, commercial and transport units

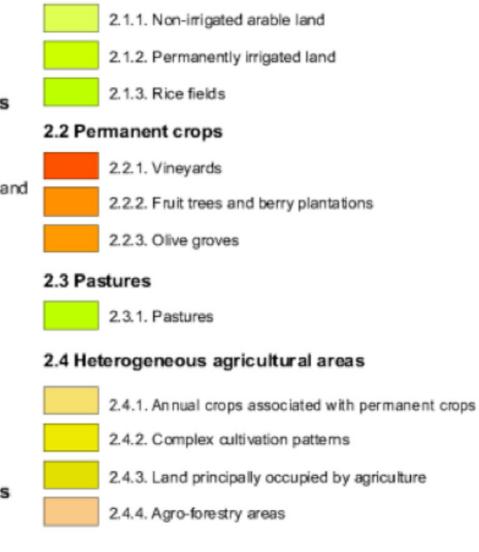
- 1.2.1. Industrial or commercial units
 1.2.2. Road and rail networks and associated land
 1.2.3. Port areas
 - 1.2.4. Airports
- 1.3 Mine, dump and construction sites
 - 1.3.1. Mineral extraction sites
 - 1.3.2. Dump sites
 - 1.3.3. Construction sites

1.4 Artificial, non-agricultural vegetated areas

1.4.1. Green urban areas 1.4.2. Sport and leisure facilities

2. Agricultural areas

2.1 Arable land



CORINE clases

Artificial surfaces Agricultural areas

CORINE clases

Forest and seminatural areas

Wetlands

Water bodies

3. Forest and seminatural areas

3.1 Forests

3.1.1. Broad-leaved forest

3.1.2. Coniferous forest

3.1.3. Mixed forest

3.2 Shrub and/or herbaceous vegetation associations

3.2.1. Natural grassland

3.2.2. Moors and heathland

3.2.3. Sclerophyllous vegetation

3.2.4. Transitional woodland shrub

3.3 Open spaces with little or no vegetation 3.3.1. Beaches, dunes, and sand plains

3.3.2. Bare rock

3.3.3. Sparsely vegetated areas 3.3.4. Burnt areas

3.3.5. Glaciers and perpetual snow

4. Wetlands

4.1 Inland wetlands



4.2 Coastal wetlands



5. Water bodies

5.1 Inland waters



5.1.1. Water courses

5.1.2. Water bodies

5.2 Marine waters





CORINE clases

Problems with

CORINE

We need to classify

Land Use classes



1.1.2. Discontinuous urban fabric



1.2.1. Industrial or commercial units



1.2.3. Port areas



1.4.1. Green urban areas

1.4.2. Sport and leisure facilities



2.1.1. Non-irrigated arable land

2.1.2. Permanently irrigated land

2.3.1. Pastures



CORINE clases

Example of **CORINE** class



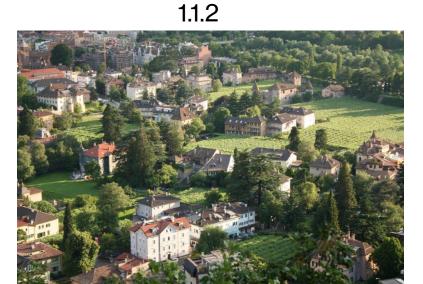
1.1.2. Discontinuous urban fabric

1.1.2



1.1.2



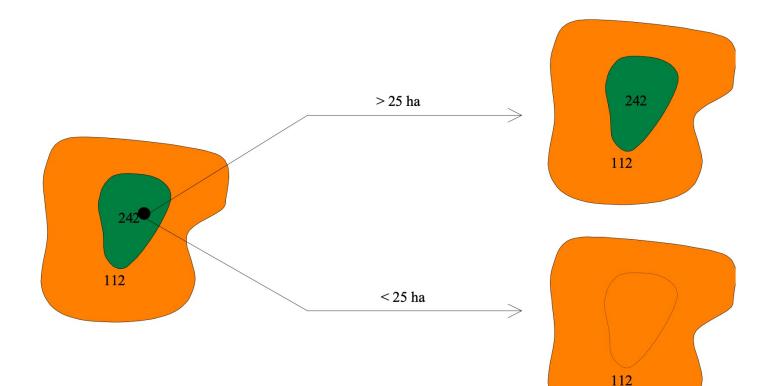


CORINE - definition and determination of classes

- Method:
- visual interpretation using remote sensing data
- smallest mapped unit is 25 ha
- Example:

 vegetated space surrounded by discontinuous urban fabric

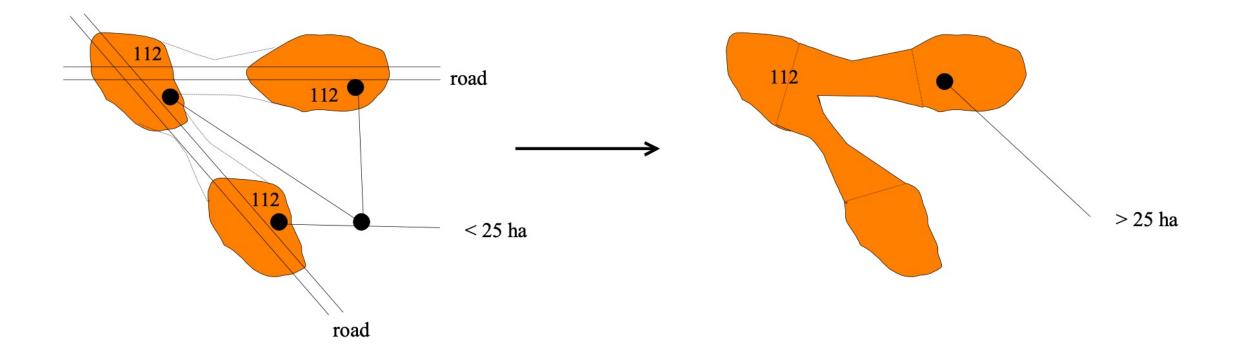
1.1.2. Discontinuous urban fabric areas



CORINE - definition and determination of classes

1.1.2. Discontinuous urban fabric areas

- grouped together if the distance < 300 m to reach 25 ha
- exterior contour line leans on road network.



Open Street Map (OSM)

- The OSMs are an iterative ongoing work-in-progress.
- The aim is to produce maps which *can* be relied upon, equally well, or better than other maps.
- OpenStreetMap is not a complete or accurate map of the world
- and should not be used in such a manner that deficiencies, omissions, inaccuracies or errors could result in death, loss or injury.
- <u>The moneclature respects keys and their values of individual</u> <u>countries</u>
- <u>Disclaimer</u>.the maps might not be reliable.



Open Street Map

- **1** Primary features
 - 1.1 Aerialway
 - 1.2 Aeroway
 - 1.3 Amenity
 - 1.3.1 Sustenance
 - 1.3.2 Education
 - 1.3.3 Transportation
 - 1.3.4 Financial
 - 1.3.5 Healthcare
 - 1.3.6 Entertainment, Arts & Culture
 - 1.3.7 Others
 - 1.4 Barrier
 - 1.4.1 Linear barriers
 - 1.4.2 Access control on highways



- 2 Additional properties
 - 2.1 Addresses
 - 2.1.1 Tags for individual houses
 - 2.1.2 For countries using hamlet, subdistrict, district, province, state
 - 2.1.3 Tags for interpolation ways
 - 2.2 Annotation
 - 2.3 Name
 - 2.4 Properties
 - 2.5 References
 - 2.6 Restrictions

Open Street Map

1. Primary features

1.1. Aerialway

This is used to tag different forms of transportation for people or goods by using aerial wires. For example these may include cable-cars, chair-lifts and drag-lifts. See the page **Aerialway** for more information on the usage of these tags.

Кеу	Value	Element	Description	Map rendering	Image	Count
aerialway	cable_car		A cable car run. Just one or two large cars. The traction cable forms a loop, but the cars do not loop around, they just move up and down on their own side, rolling along static cables over which they are suspended.	••••	A	3 • 1 337 💽 11 💽
aerialway	gondola		An aerialway where the cabins go around in a circle	••••		1 ⊙ 1 543 ∢ 30 💽

Open Street Map



1 Primary features 1.1 Aerialway 1.2 Aeroway 1.3 Amenity 1.3.1 Sustenance

Amenity

Used to map facilities used by visitors and residents. For example: toilets, telephones, banks, pharmacies, cafes, parking and schools. See the page **Amenities** for an introduction on its usage.

Кеу	Value	Element	Comment	carto-Rendering 🗗	Photo
amenity	bar	•	Bar is a purpose-built commercial establishment that sells alcoholic drinks to be consumed on the premises. They are characterised by a noisy and vibrant atmosphere, similar to a party and usually don't sell food. See also the description of the tags amenity=pub; bar; restaurant for a distinction between these.	X	

CORINE versus

Open Street Map

CORINE	OSM		
	nodes - defining points in space		
polygons	ways - defining linear features and area boundaries		
polygons			
3 level categories			
1 st 5 classes	29 Primary Features of the 1 st hierarchy levels		
2 nd 15 classes	2 levels of Primary Features		
3 rd 44 classes	(202 values of landuse in CR)		
Updated CLC illustrated nomenclature guidelines, Kosztra, B. et al.	https://wiki.openstreetmap.org/wiki/Map_features		

Conversion between

Open Street Map (202 tags in CR) and **CORINE** (2nd level 15 classes)

CORINE

OSM	-	• • • • •	_													
	11	12	13	14	21	22	23	24	31	32	33	41	42	51	52	2
	Urban f	Industrial,	Mine, dump	Artificial, non-	Arable land	Permanent	Pastures	Heterogeneous	Forests	Scrub and/o	Open space	Inland wetla	Maritime w	Inland waters	Marine waters	Others
landuse																
allotments	0	0	0	1	0	0	0	0	0	0	0	0	0	0	0)
basin	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0)
brownfield	0	0	1	0	0	0	0	0	0	0	0	0	0	0	0)
cemetery	0	1	0	0	0	0	0	0	0	0	0	0	0	0	0)
commercial	0	1	0	0	0	0	0	0	0	0	0	0	0	0	0)
construction	0	0	1	0	0	0	0	0	0	0	0	0	0	0	0	ז 🖌
farmland	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0)
farmyard	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0)
forest	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	ז
garages	0	1	0	0	0	0	0	0	0	0	0	0	0	0	0	ז
grass	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	2
greenfield	0	0	1	0	0	0	0	0	0	0	0	0	0	0	0)
greenhouse_hortic ulture	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0)
industrial	0	1	0	0		0	-	-	0	0			-	-	0	5
landfill	0	0	1	0		0	-		0	0				0	0	5
meadow	0	0	0	0		0	1	0	0					0	0	5
orchard	0	0	0	0		1	0	0	0	0		64.		0	0	<u>,</u>
plant_nursery	0	0	0	0		1	0		0						0	5
quarry	0	0	1	0	-	0	-		0	0				0	0	5
railway	0	1	0	0	200	0	0		0			1995	1000		0	<u>,</u>
recreation_ground	0	0	0	1	0	0	600		0			C4.			0	5
reservoir	0	0	0	0	0	0			-	-					0)

LUCAS Land Use/Cover Area frame Survey



"Point data"at a systematic grid of 2 x 2 km

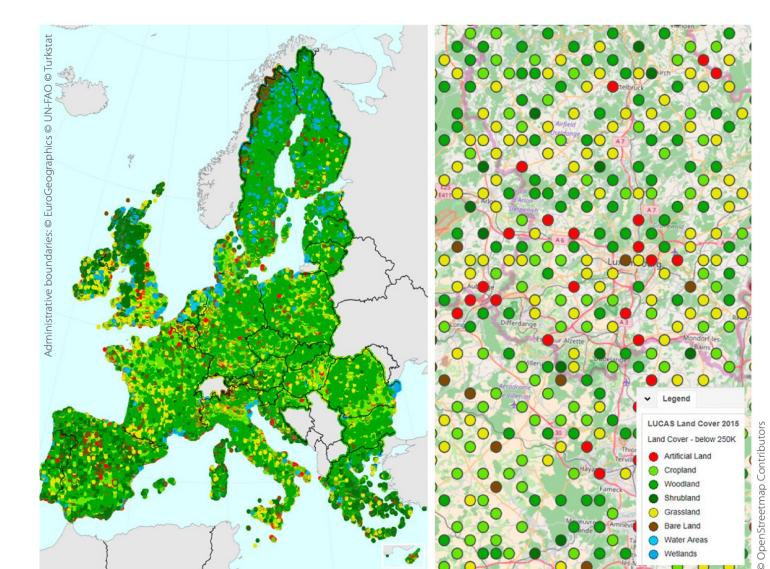
Years: 2005, 2009, 2012, 2015, 2018

2015 over 270 000 points; **2018** 1 090 863 points

Method

- field observation and
- photo-interpretation

ETRS89-extended / LAEA Europe as spatial reference system (EPSG:3035)



Artificial land

Cropland









LUCAS Land Cover

1st level 8 categories

2nd level 30 categories

3rd level 76 categories

Land	cover			Artificial land	Cropland
A00	ARTIFICIAL LAND	A10	Roofed built-up areas		
		A20	Artificial non-built up areas		
		A30	Other artificial areas	A AND THE ALL	
B00	CROPLAND	B10	Cereals		
		B20	Root crops		ALL
		B30	Non-permanent industrial crops	© European Union LUCAS	PEuropean Union LUCAS
		B40	Dry pulses, vegetables and flowers	Woodland	Shrubland
		B50	Fodder crops		
		B70	Permanent crops: fruit trees	一种 医闭口 医门口的	
		B80	Other permanent crops		and here and
C00	WOODLAND	C10	Broadleaved woodland		a sea a fair
		C20	Coniferous woodland	WW BRIDDENIA PURCHARMAN	
		C30	Mixed woodland		O Furopean Union
D00	SHRUBLAND	D10	Shrubland with sparse tree cover	European Union LUCAS	LUCAS
		D20	Shrubland without tree cover	Grassland	Bare land
E00	GRASSLAND	E10	Grassland with sparse tree/shrub cover		
		E20	Grassland without tree/shrub cover		
		E30	Spontaneously re-vegetated surfaces		
F00	BARE LAND AND LICHENS/MOSS	F10	Rocks and stones	A REAL PROPERTY AND A REAL	
		F20	Sand		
		F30	Lichens and moss	European Union UUCAS	© European Union
		F40	Other bare soil	LUCAS	LUCAS
G00	WATER AREAS	G10	Inland water bodies	Water areas	Wetlands
		G20	Inland running water	The second s	
		G30	Transitional water bodies		
		G40	Sea and ocean		
		G50	Glaciers, permanent snow		and the second second
H00	WETLANDS	H10	Inland wetlands		1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
		H20	Coastal wetlands	© European Union	Pivabay

LUCAS Land Cover Categories

3rd level 76 subclasses

3 levels of the Artificial Land

A00 ARTIFICIAL LAND A10 BUILT-UP AREAS A11 Buildings with one to three floors A12 Buildings with more than three floors A13 Greenhouses A20 ARTIFICIAL NON-BUILT UP AREAS A21 Non built-up area features A22 Non built-up linear features

LUCAS

Land Use Categories

4 sectors 33 subclasses 2 levels of Land Use - example of the secondary sector

U220 INDUSTRY AND MANUFACTURING

U221 Manufacturing of food, beverages and tobacco products U222 Manufacturing of textile products U223 Coal, oil and metal processing U224 Production of Non-metal mineral goods U225 Chemical and allied industries and manufacturing U226 Machinery and equipment U227 Wood based products



LUCAS - link between LC and LU

FOO BARE LAND AND LICHENS/MOSS

Land Cover

links to

Land Use

F10 ROCKS AND STONES

- This class includes
 - Inland rock cliffs
 - Areas of rock outcrop and limestone parent
 - Scree
 - Block litter and mountain top debris
 - Land covered with recent volcanic features

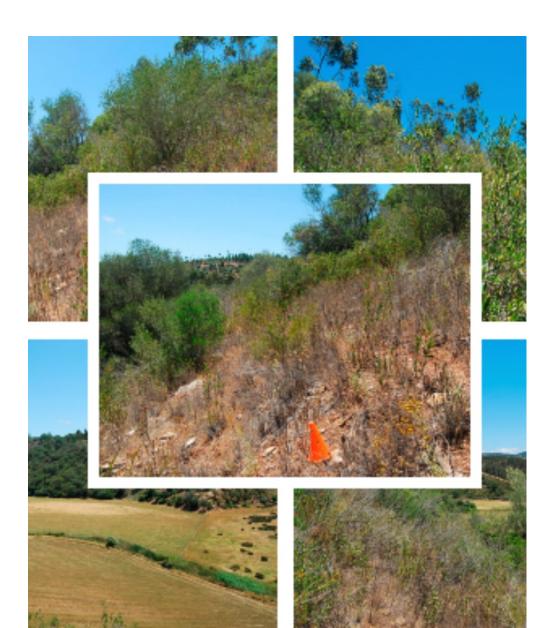
- F10 > U140 Scrapped or excavated areas for quarrying, mining purposes
- F10 > U21x Rocky areas attached to energy production
- F10 \geq U22x Rocky areas attached to industrial use
- F10 > U31x Rocky areas attached to transport areas
- F10 ➤ U330 Construction sites
- F10 > U34x Rocky areas attached to commercial services
- F10 \geq U350 Rocky areas attached to community services
- F10 \geq U36x Rocky areas used for recreational and leisure purposes
- F10 \geq U370 Rocky areas attached to residential areas
- F10 > U4x0 Rocky areas not used (inland rocks, top mountains)

LUCAS

Method of data collection

Homogeneous land cover

point "area" (7 m²), i.e. 1,5 m radius



LUCAS

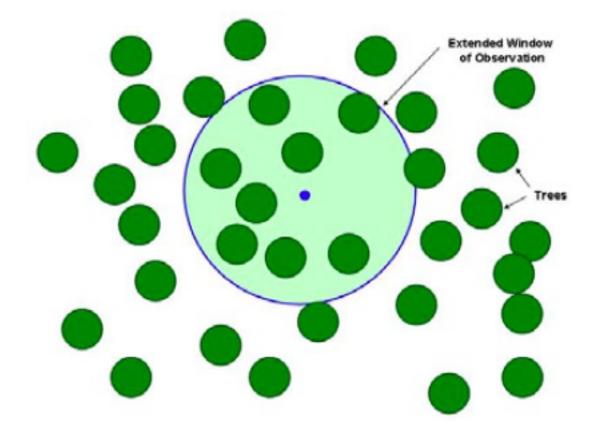
Method of data collection

Non-homogeneous land cover:

i.e., trees or shrubs interspersed with grassland,

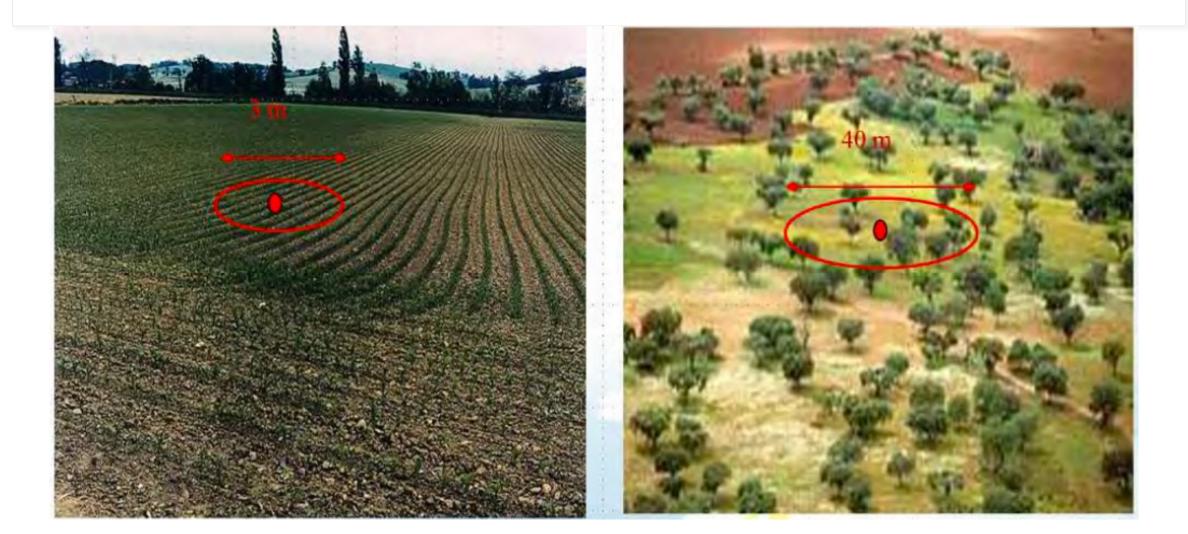
Extended Window of Observation:

a radius of 20 meters, i.e., 0,13 ha



Radius 1,5 m

Radius 20 m



Conversion between

LUCAS (30 classes 2nd level) and CORINE (to 2nd level and 3rd level classes)

LUCAS 2nd level (LC2)	LC2 code	CORINE 2nd level code		note	
Roofed built-up areas	A10				
Artificial non-built up areas	A20	12	122		
Other artificial areas	A30			could be 122 or 132	
Cereals	B10	21	211	contains rice	
Root crops	B20	21	211		
Non-permanent industrial crops	B30	21	211		
Dry pulses, vegetables and flowers	B40	21	211		
Fodder crops	B50	21	211	possible pastures (2	3)
Permanent crops: fruit trees	B70	22	222		
Other permanent crops	B80	22		vineyards, olive grov	es, nurseries?
Broadleaved woodland	C10	31	311		
Coniferous woodland	C20	31	312		
Mixed woodland	C30	31	313		

Conversion between OSM and LUCAS for the project

LUCAS

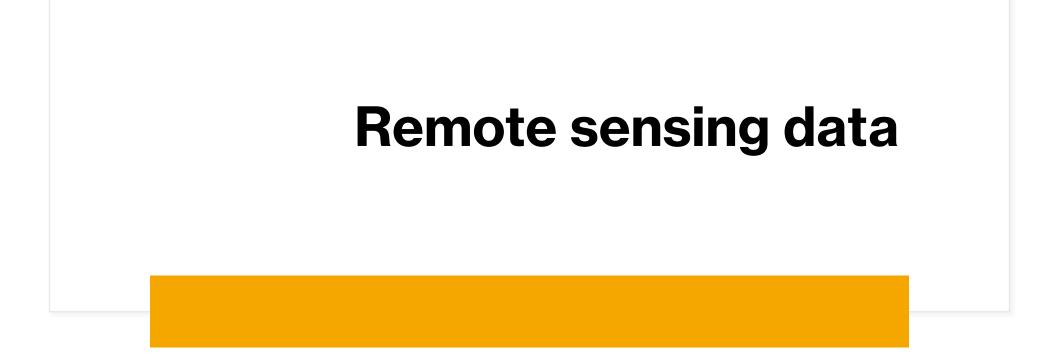
OSM 10 Classes. - LUCAS

В

OSM 20 classes - LUCAS

Land	cover					
A00	ARTIFICIAL LAND	A10	Roofed built-up areas			
		A20	Artificial non-built up areas			
		A30	Other artificial areas			
B00	CROPLAND	B10	Cereals			
		B20	Root crops			
		B30	Non-permanent industrial crops			
		B40	 Artificial non-built up areas Other artificial areas Cereals Root crops Non-permanent industrial crops Dry pulses, vegetables and flowers Fodder crops Fodder crops Permanent crops: fruit trees Other permanent crops Broadleaved woodland Coniferous woodland Mixed woodland Shrubland with sparse tree cover Shrubland without tree/shrub cover Grassland without tree/shrub cover Spontaneously re-vegetated surfaces Rocks and stones Sand Lichens and moss Other bare soil Inland water bodies Transitional water bodies Sea and ocean 			
		B50	A20Artificial non-built up areasA30Other artificial areasB10CerealsB20Root cropsB30Non-permanent industrial cropsB40Dry pulses, vegetables and flowersB50Fodder cropsB70Permanent crops: fruit treesB80Other permanent cropsC10Broadleaved woodlandC20Coniferous woodlandC30Mixed woodlandC30Shrubland with sparse tree coverB20Shrubland with sparse tree/shrub coverE30Spontaneously re-vegetated surfacesF10Rocks and stonesF20SandF30Lichens and mossF40Other bare soilG10Inland water bodiesG30Transitional water bodiesG40Sea and oceanG50Glaciers, permanent snowH10Inland wetlands			
		B70	 Fodder crops Permanent crops: fruit trees Other permanent crops Broadleaved woodland Coniferous woodland Mixed woodland Shrubland with sparse tree cover Shrubland without tree cover Grassland with sparse tree/shrub cover Grassland without tree/shrub cover 			
		B80	Other permanent crops			
C00	WOODLAND	C10	Broadleaved woodland			
			Coniferous woodland			
		C30	Mixed woodland			
D00	SHRUBLAND	D10	Shrubland with sparse tree cover			
		D20	Shrubland without tree cover			
E00	GRASSLAND	E10	Grassland with sparse tree/shrub cover			
		E20	Grassland without tree/shrub cover			
		E30	Spontaneously re-vegetated surfaces			
F00	BARE LAND AND LICHENS/MOSS	F10	Rocks and stones			
		F20	Sand			
		F30	 Mixed woodland Shrubland with sparse tree cover Shrubland without tree cover Grassland with sparse tree/shrub cover Grassland without tree/shrub cover Grassland without tree/shrub cover Spontaneously re-vegetated surfaces Rocks and stones Rocks and stones Sand Lichens and moss Other bare soil Inland water bodies Inland running water 			
		F40	Other bare soil			
G00	WATER AREAS	G10	Inland water bodies			
		G20	Inland running water			
		G30	Transitional water bodies			
		G40	Sea and ocean			
		G50	Glaciers, permanent snow			
H00	WETLANDS	H10	Inland wetlands			
		H20	Coastal wetlands			

Artificial	A10, A20	Roofed buit-up areas	A10
		Artificial non-built up	A20
		Mine, dump	
		Urban green	
Cropland	B10B40	Cropland seasonal	B10B40
	B50	Irrigated cropland	
		Pastures	B50?
Perenial crops	B70, B80	Perenial crops	B70, B80?
orest	C10, C20	Forest Broadleaf	C10
		Forest Coniferous	C20
		Forest Mixed	C30
Shrubland	D20	Shrubland	D20
Grassland	E20, B50?	Natural grassland	E20
Barren	F00	Barren	F00
Vetlands	H00	Inland Wetlands	H10
		Coastal wetlands	H20
Vater	G10, G20, G40	Water bodies	G10
		Water courses	G20
		Ocean	G40
Glaciers, permanent snow	G50	Glaciers, permanent snow	G50



Open data: Landsat, Sentinel Data

2000 - 2019

Landsat data – Landsat 5 (TM), Landsat 7 (ETM+), Landsat 8 (OLI)

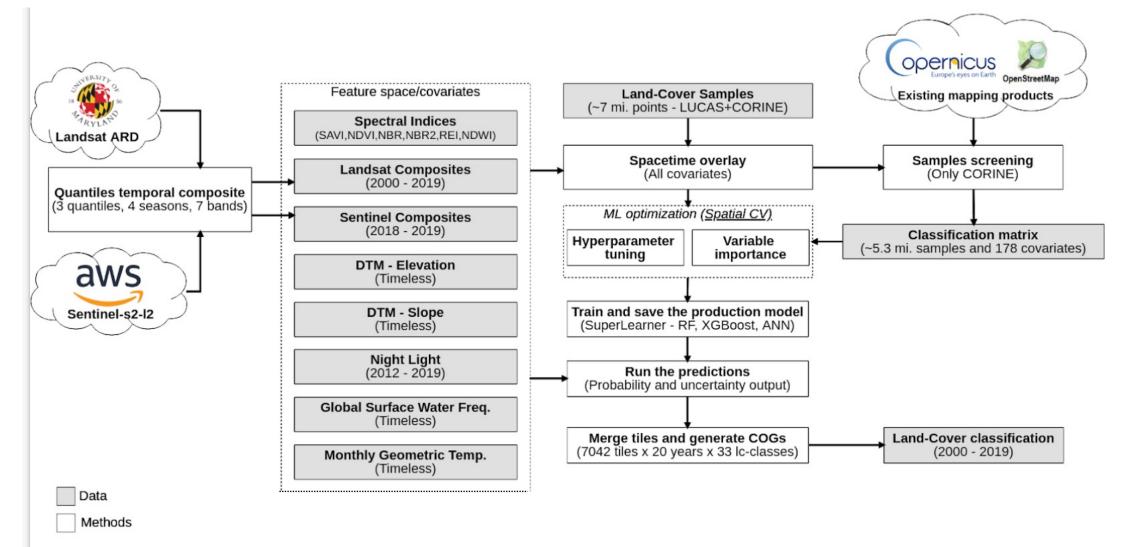
2015 - 2019

Sentinel-2

4 seasons :

Winter: December 2 of previous year until March 20 of current year
Spring: March 21 until June 24 of current year
Summer: June 25 until September 12 of current year
Fall: September 13 until December 1 of current year

EU land cover mapping - workflow



EU land cover mapping: workflow

satellite data preprocessing

Landsat Composites (2000 - 2019)

Sentinel Composites (2018 - 2019)

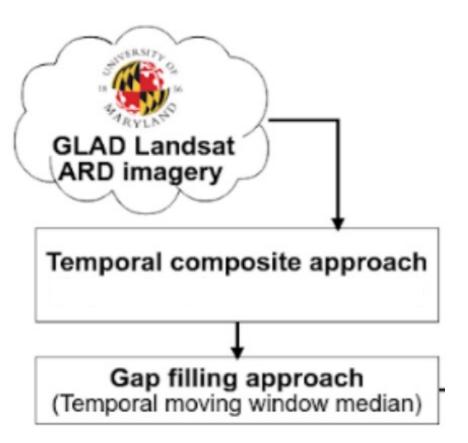
> Night Light (2012 - 2019)

Night Light / VIIRS/NPP

Visible Infrared Imaging Radiometer Suite (VIIRS/NPP) is nighttime radiance product

EU land cover mapping: workflow

1st step satellite data preprocessing gap filling



Gap Filling for Landsat Mosaics

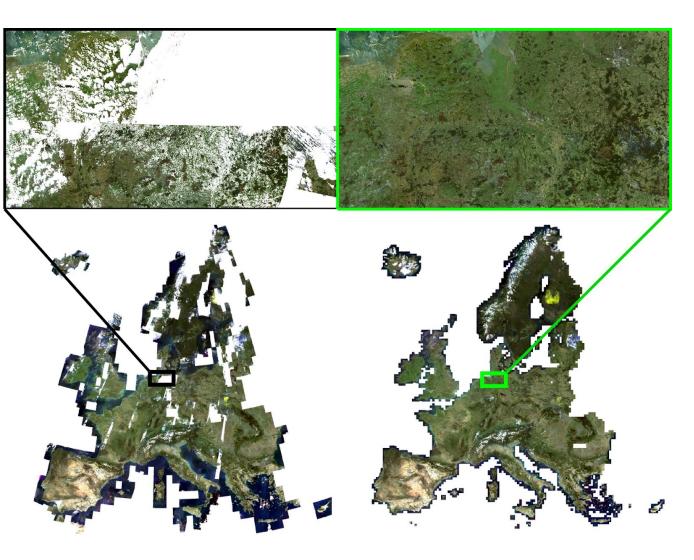
Where: clouds, shadows, non-existing data,

Method: based on Landsat Analysis Ready Data (GLAD ARD) developed by the <u>Global Land Analysis and Discovery</u> (GLAD)

(Landsat Analysis Ready Data for Global Land Cover and Land Cover Change Mapping Peter Potapov *, Matthew C. Hansen , Indrani Kommareddy, Anil Kommareddy, Svetlana Turubanova, Amy Pickens, Bernard Adusei, Alexandra Tyukavina and Qing Ying Remote Sens. 2020, 12, 426; doi:10.3390/rs12030426)



Gap Filling for Landsat Mosaics



Method: Temporal Moving Window Median-TMWM temporal_step=8

Production:

All the <u>GeoHarmonizer tiles</u> were gapfilled generating 23,661,120 individual files,

<u>7 bands</u> (Blue, Green, Red, NIR, SWIR1, SWIR2 and Thermal),

<u>20 years</u> (2000 - 2019) and <u>4 seasons</u> were processed.

Gap Filling for Landsat Mosaics

Long-Term Median Improvement

Pixels with gaps are filled in by the following options

9 priorities combined with 3 different sizes

1st – 3rd priority = windows 1, 2 3 from **the same season**: *median value* 4th – 6th priority = windows 1, 2, 3 from **fall and spring**: *average of medians* 7th – 9th priority = windows 1, 2, 3 from **spring**, **summer and fall**: *median*

3 sizes of moving windows:

window 1: 8 neighboring periods - 4 before and after window 2: 16 neighboring periods – 8 before and after window 3: 24 neighboring periods – 12 before and after

The 1st preference the same season with window 1

EU land cover mapping: workflow

Multispectral image enhancement

Spectral Indices (SAVI,NDVI,NBR,NBR2,REI,NDWI)

Soil Adjusted Vegetation Index

$$SAVI = \frac{1.5 * (NIR - Red)}{(NIR + Red + 0.5)}$$

 $NDVI = \frac{(NIR - Red)}{(NIR + Red)}$

Normalized Burn Ratio Riparian Ecosystem

NBR = (NIR - SWIR) / (NIR + SWIR)NBR2 = (SWIR1 - SWIR2) / (SWIR1 + SWIR2)

Riparian Ecosystem Index

Normalized Difference Water Index $REI = NIR2 - B/NIR2 + B \times NIR2$

 $NDWI = \frac{(\rho_{857} - \rho_{1241})}{(\rho_{857} + \rho_{1241})}$

EU land cover mapping workflow

 Global Surface Water Freq. (Timeless)

 Other data layers
 Monthly Geometric Temp. (Timeless)

 Global Surface Water

Value	Label
0	Not water
1	1% occurrence
100	100 % occurrence
255	No data

Global Surface Water

1984 – 2019 JRC: https://global-surface-water.appspot.com/download

Monthly Geometric Temperature

DTM - Elevation (Timeless)

DTM - Slope
(Timeless)

Spatio-temporal interpolation of daily temperatures for global land areas at 1km resolution Milan Kilibarda et al

https://agupubs.onlinelibrary.wiley.com/doi/full/10.1002/2013JD020803

EU land cover mapping: workflow

training and testing pixel selection



Training data	Validation data
	Lucas (20%)
Lucas (80%)	Corine (100%)
	Corine (20%)
Corine (80%)	Lucas (100%)
All (80%)	All (20%)

7 million pixels selected 5,3 mi used for classification

EU land cover mapping – workflow

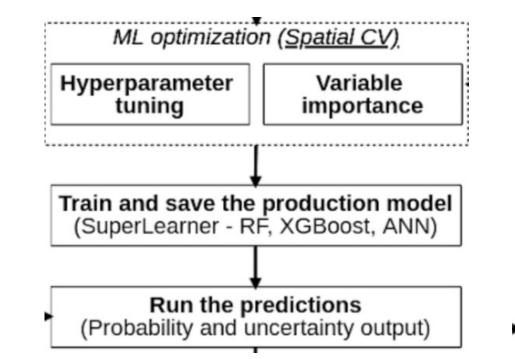
training and testing pixels selection

Improvement of selection of pixels

lc_class	condition	tree_cover	grasslands	imperv
111: Urban fabric	AND			
122: Road and rail networks and associated land	OR			>
131: Mineral extraction sites	AND	equals 0	equals 0	
141: Green urban areas	OR	> 0	> 0	
211: Non-irrigated arable land	AND	equals 0		
221: Vineyards	AND		equals 0	
222: Fruit trees and berry plantations	AND		equals 0	
223: Olive groves	AND		equals 0	

EU land cover mapping – workflow

classification



Classification matrix 5,3 mi pixels 178 covariates



(2000 - 2019)

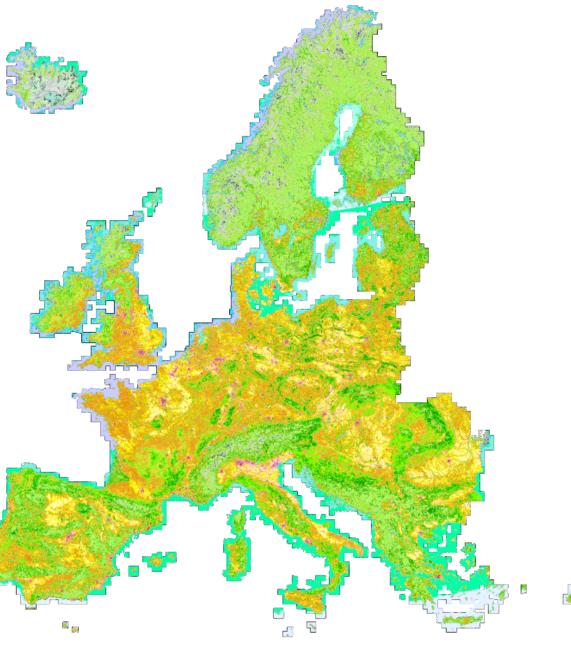
111-Urban fabric 122-Road and rail networks and associated land 123-Port areas 124-Airports 131-Mineral extraction sites 132-Dump sites 133-Construction sites 141 Green urban areas 211-Non-irrigated arable land 212-Permanently irrigated arable land 213-Rice fields 221-Vineyards 222-Fruit trees and berry plantations 223-Olive groves 231-Pastures 311-Broad-leaved forest 312-Coniferous forest 321-Natural grasslands 322-Moors and heathland 323-Sclerophyllous vegetation 324-Transitional woodland-shrub 331-Beaches, dunes, sands 332-Bare rocks 333-Sparsely vegetated areas 334-Burnt areas 335-Glaciers and perpetual snow 411-Inland wetlands 421-Maritime wetlands 511-Water courses 512-Water bodies 521-Coastal lagoons 522-Estuaries

523-Sea and ocean



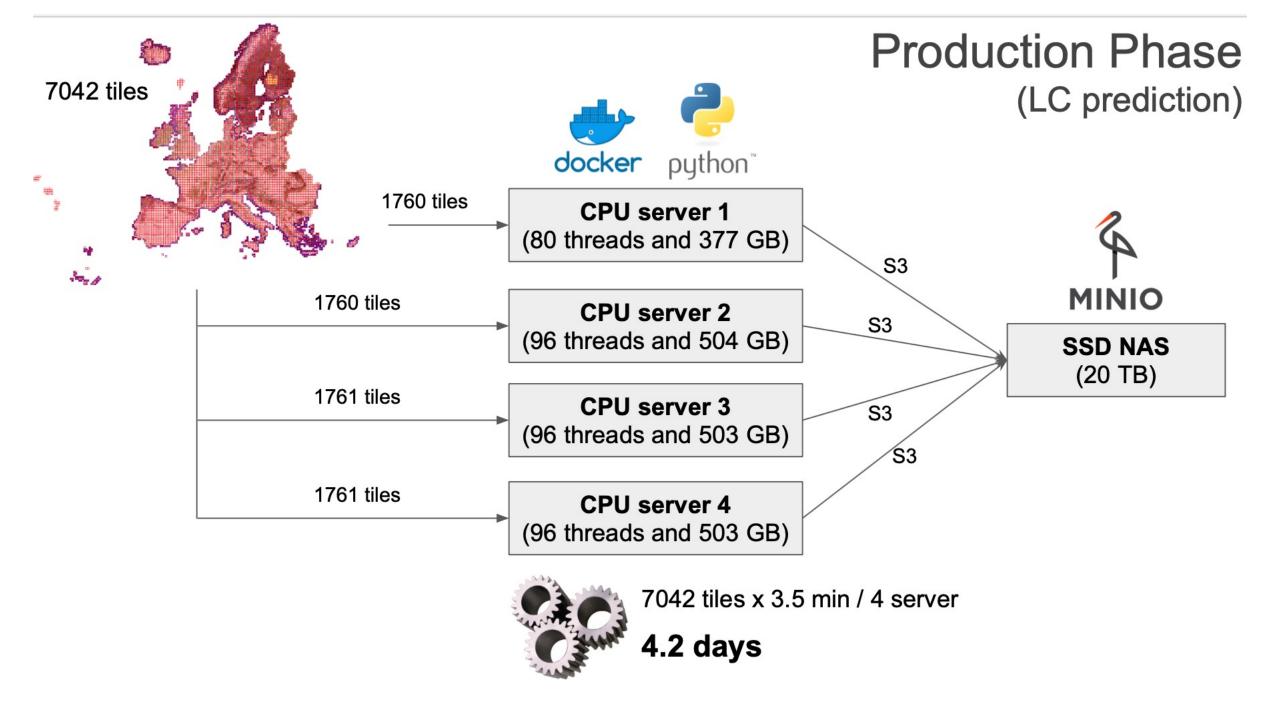
Unclassified CORINE classes:

112 Discontinuos urban fabric 121 Industrial or commercial units 142 Sport and leisure activities 24 Heterogenous agricultural areas 313 Mixed forest 412 Peat bogs 422 Salines 423 Inertidal flats





С¦Г



Conclusion

Land Cover, its classification and harmonization is a complex problem due to:

1) Existing detailed namenclatures and their harmonization

- a. which are a combination of LU and LC
- b. and therefore the class definitions mostly do not respect spectral feature space
- b. where many classes in one nomenclature comprise more classes of the other nomenclature

2) Atmospheric conditions – even though gap filling methods are able to replace non-existing data, the training phase uses many pixels where the spectral values is a combination of spectral values even of all seasons.

Conclusion

The project is based on **open data** – therefore we are limited in the remote sensing data choice which means a relatively low number of remote sensing data especially before Sentinel-2

The rest of the project will be focused on

the nomenclature tuning and

training of the machine learning models to improve the accuracy which has not yet reached acceptable values.

The project ends in June 2022