

# **Surface parameters evaluated from satellite remote sensing images for pollutant atmospheric dispersion modelling**

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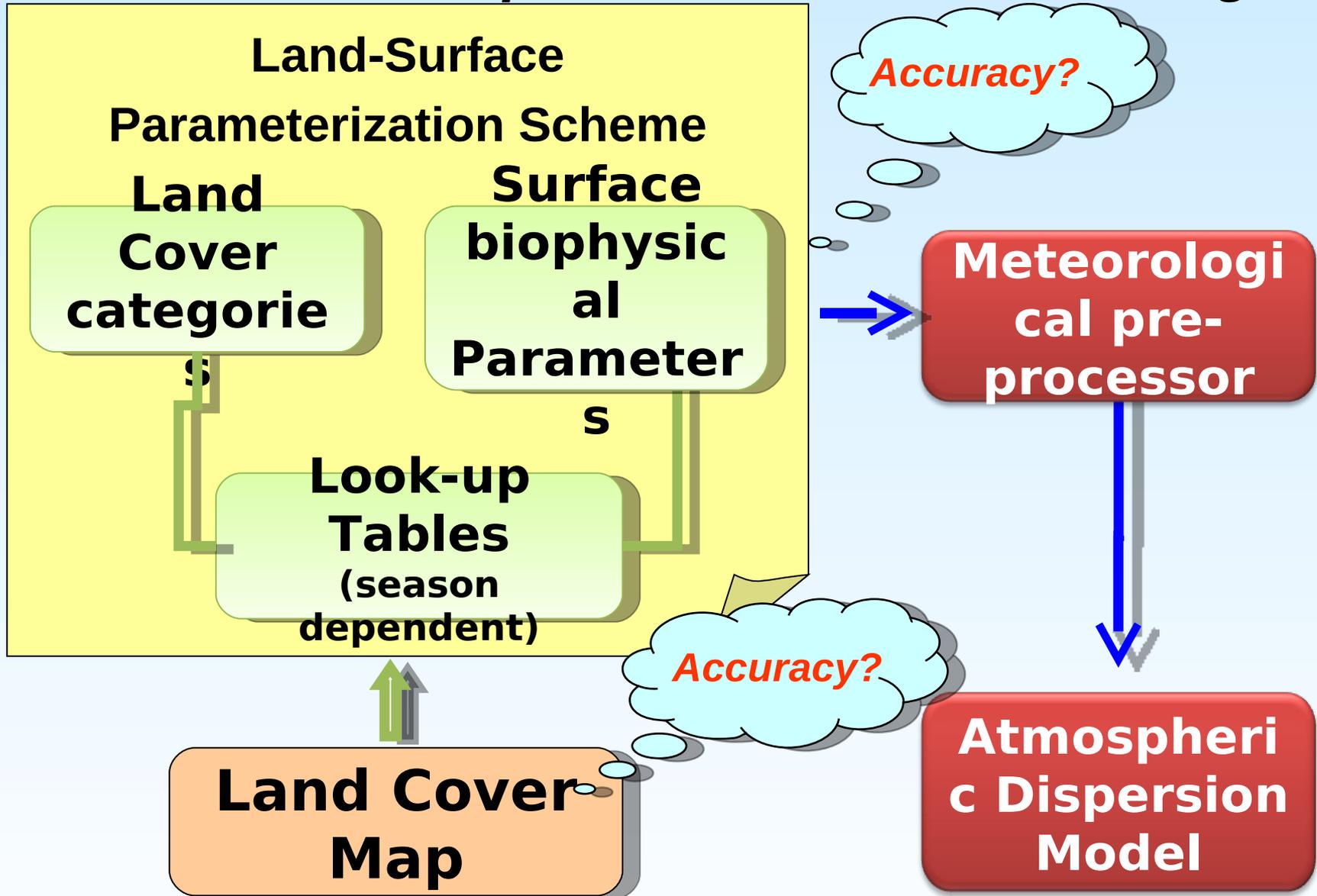


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# Surface-Atmosphere Interaction Modelling



## ***Work Rationale***

Satellite remote sensing can contribute to increase accuracy by:

- updating the land cover maps,
- directly estimating some surface biophysical parameters.

**But WE NEED**

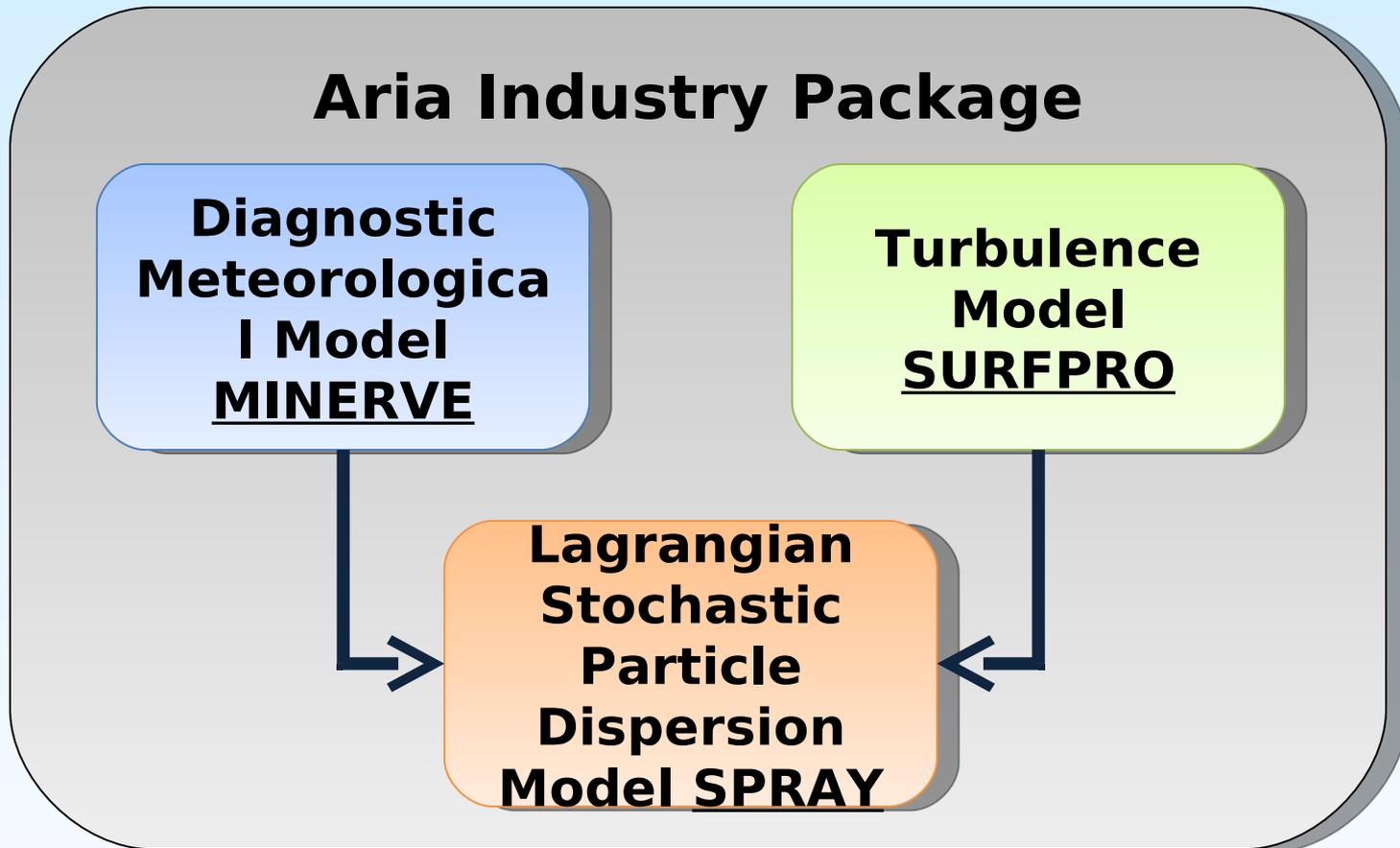
- operational methodologies of data processing;
- suitable interfaces to feed atmospheric models

## ***Work Summary***

We simulated the dispersion of the pollutants emitted from industrial sources using ASTER satellite images to feed the model with:

- updated CORINE Land Cover map
- calculated surface albedo map

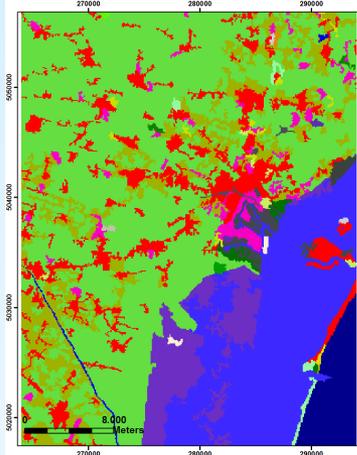
# ***Atmospheric Dispersion Model***



*Tinarelli G. et al., 1998; Geai P., 1987; ARIANet R2006.23, 2006*

# ***Land-Surface Parameterization in the SURFPRO model***

## ***Land Cover Land Use map***



***(2D array)***

**LOOK-UP  
Tables**  
(season dependent)

### ***Surface Parameters:***

- Roughness length
- Albedo
- Bowen ratio
- Soil heat flux
- Leaf area index
- Canopy height
- Internal resistance

The ARIA Industry Package includes **2** parameterization schemes:

**Aggregate CORINE Scheme** and **BATS Scheme**

CORINE Legend - Level 3	New Aggregated CORINE Scheme
Continuous urban fabric	Continuous urban fabric
Discontinuous urban fabric	Discontinuous urban fabric
Industrial or commercial units	Industrial
Port areas	
Road and rail networks and associated land	Open, mostly artificial surfaces
Airports	
Construction sites	
Dump sites	
Non-irrigated arable land	Cultivated land
Permanently irrigated land	
Rice fields	Rice fields
Vineyards	Permanent/woody crops
Fruit trees and berry plantations	
Olive groves	
Pastures	Pastures
Annual crops associated with permanent crops	Crops/wood mosaic
Land principally occupied by agricult., with signific. areas of natural	
Green urban areas	
Sport and leisure facilities	
Complex cultivation patterns	Scattered settlement (farms, villages, trees, hedge
Broad-leaved forest	Broad-leaved forest
Coniferous forest	Coniferous forest
Mixed forest	Mixed forest
Agro-forestry areas	
Moors and heathland	Deciduous or evergreen Bush/Shrub vegetation
Sclerophyllous vegetation	
Transitional woodland-shrub	
Beaches, dunes, and sand plains	Beaches, dunes, and sand plains
Bare rock	Bare rock
Mineral extraction sites	Sparsely vegetated areas
Sparsely vegetated areas	
Burnt areas	
Natural grassland	
Glaciers and perpetual snow	Glaciers and perpetual snow
Inland marshes	Inland and coastal wetlands
Peatbogs	
Salt marshes	
Salines	
Intertidal flats	
Water courses	Water bodies
Water bodies	
Coastal lagoons	
Estuaries	
Sea and ocean	

## Aggregated CORINE scheme

Classes added for remote sensing

based on the European  
CORINE and Cover  
classification

Cultivated land: Fallow ground/Stubbles	(FA)
Cultivated land: Very young crops	(SP)
Cultivated land: Low/mature green crops	(SU)
Cultivated land: High senescent crops	(SU)
Cultivated land - generic	
Pastures - generic	
Pastures: Thin sparse grass	
Pastures: Thick/homogeneous grass	(SU)

21 land use categories obtained  
grouping the 44 classes of the  
CORINE Level 3 legend

Parameters were assigned  
using values of the original  
class, for the season when the  
considered phenological  
Slightly modified by the authors  
condition is predominant

used to run the model  
using the Regional  
Sectoral Updated  
maps

# Simulation Domain and Model Setup

- 30 km x 40 km domain
- Part of the Venice lagoon (N Italy)
- Porto Marghera industrial site
- Almost flat coastal area
- Sea-breeze development
- Complex land use pattern (urban and industrial areas, cultivated land, sea, lagoon, marsh areas)

~~Emission sources: 2 virtual stacks~~

~~Time step: 1 hour 100 m~~

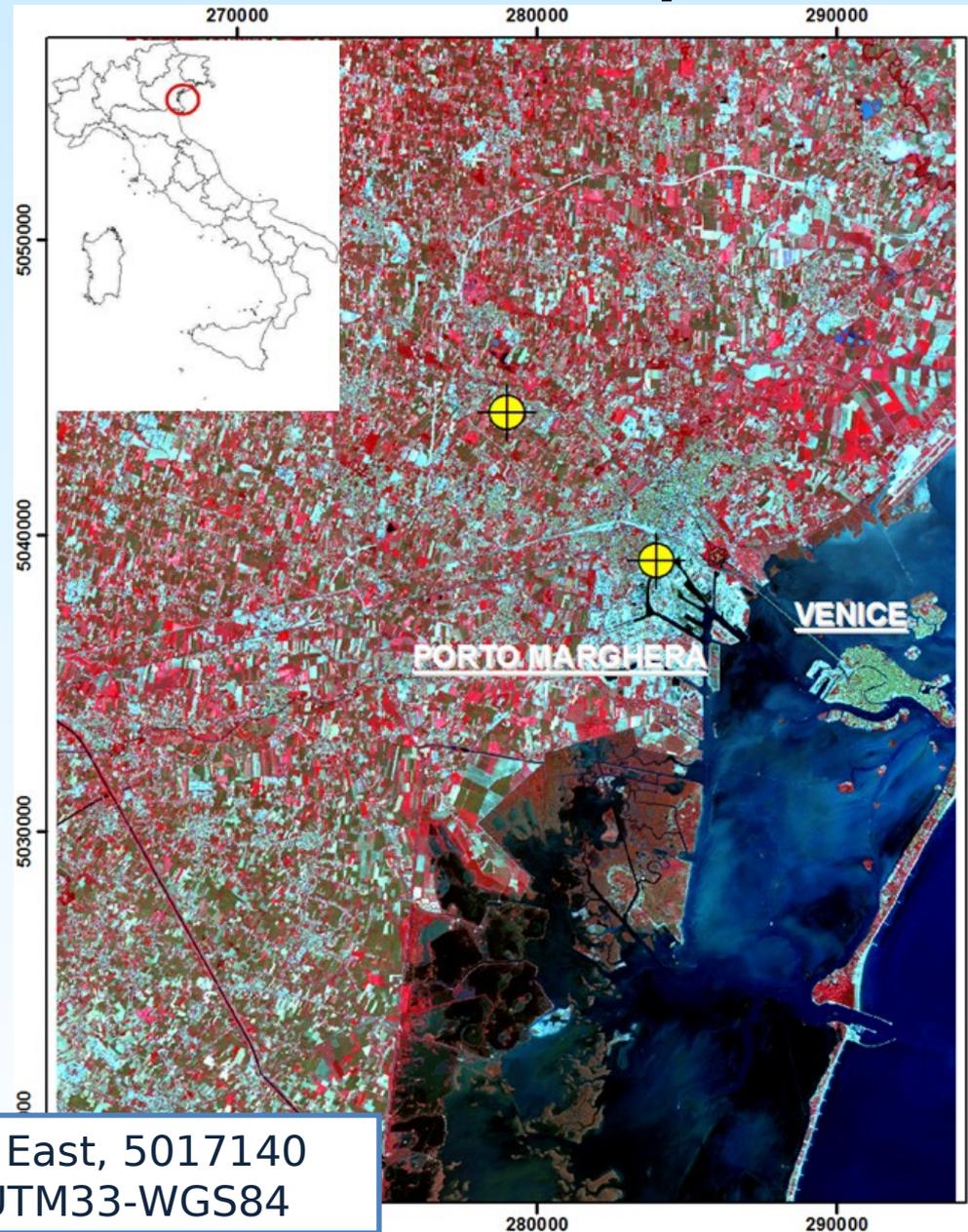
~~Diameter: 4 m~~

~~Spatial domain of the 3D grid: Temperature of gases: 140 °C~~

~~Horizontal: 121 x 161 nodes, step = 250 m;~~

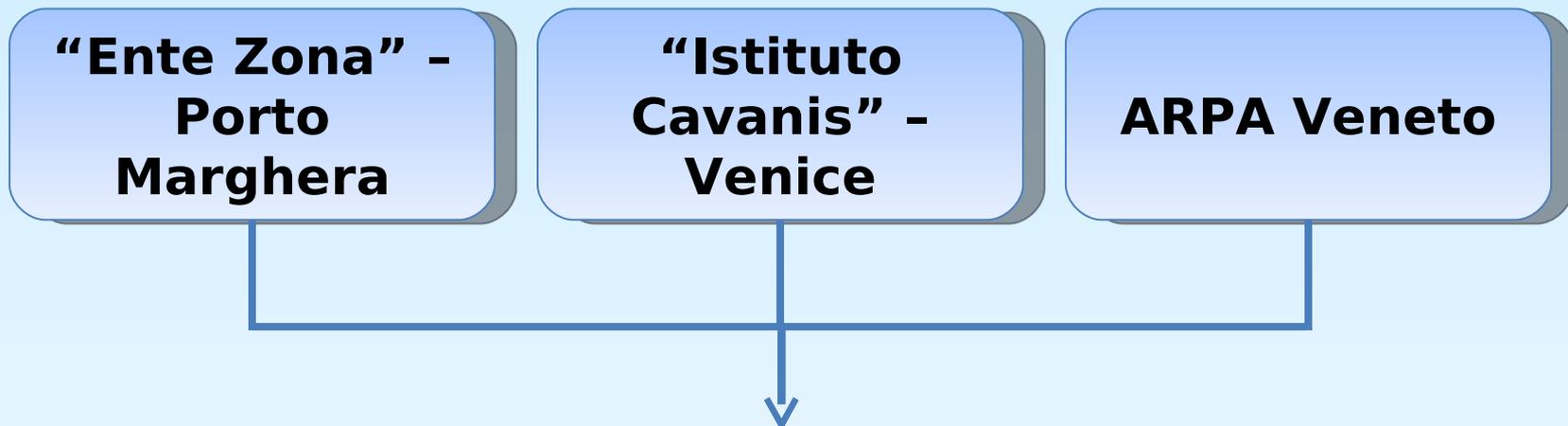
~~Emission: 3240  
Vertical: 30 nodes, from ground to 1500 m,  
particles h-1~~

variable step starting from 10 m

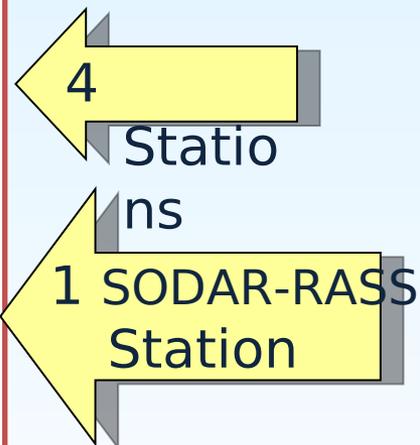


264000 m East, 5017140  
m North UTM33-WGS84

# Meteorological Dataset

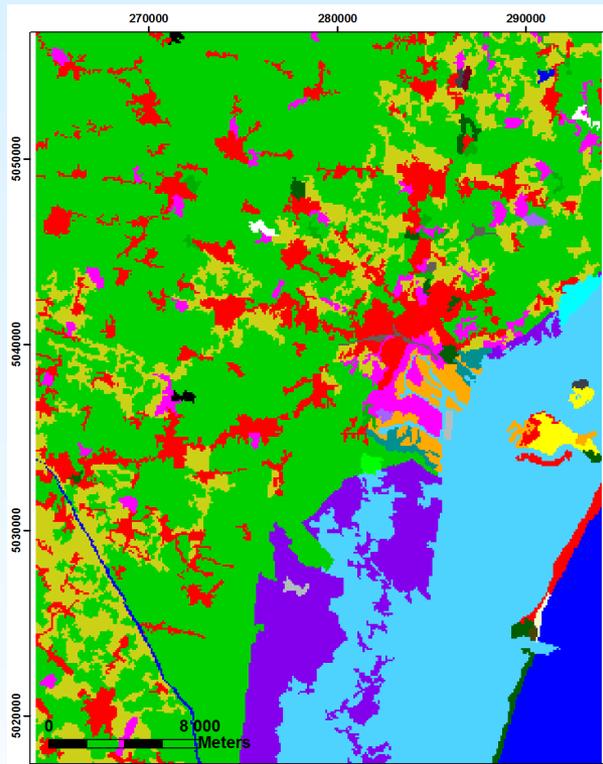


- Hourly time series of: Wind speed and direction, Air Temperature, Air Pressure, Relative Humidity, Solar radiation
- Atmospheric Vertical Profiles of Wind speed and direction and Air Temperature
- Water (Sea, lagoon) Temperature



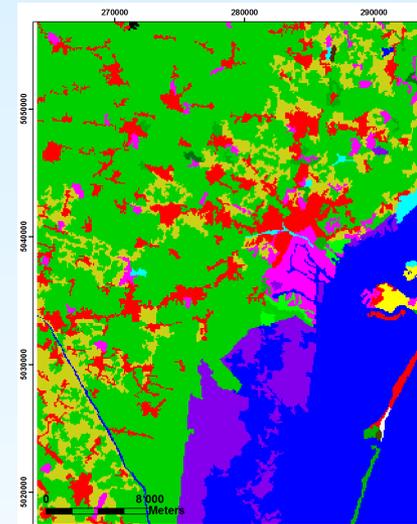
# Land Cover map

**CLC2000 CORINE map – Level 3**

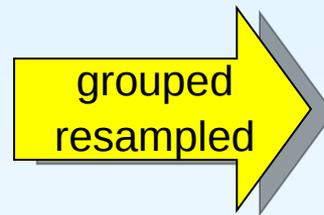


Resolution: 100 m

**Aggregated CORINE map**



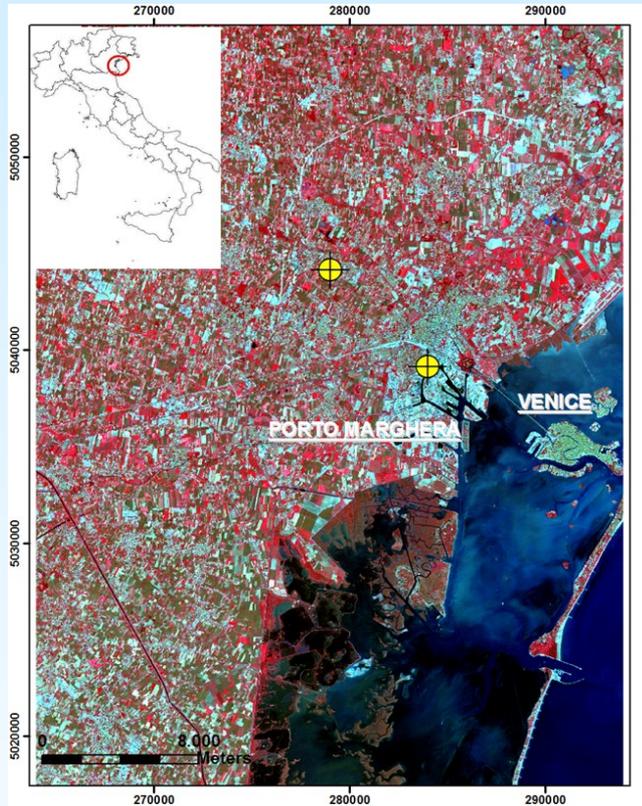
Resolution: 250 m



## Data set

# ASTER Satellite Image

Image acquired on 5 September 2007



Pre processed to obtain georeferenced images of spectral surface reflectance, corrected for atmospheric effects.

## ASTER Sensor

Band Number	Electromagnetic Range ( $\mu\text{m}$ )		Spatial Resolution (m)
1	VNIR	0.52-0.60	15
2		0.63-0.69	
3N/3B		0.78-0.86	
4	SWIR	1.60-1.70	30
5		2.145-2.185	
6		2.185-2.225	
7		2.235-2.285	
8		2.295-2.365	
9		2.360-2.430	
10	TIR	8.125-8.475	90
11		8.475-8.825	
12		8.925-9.275	
13		10.25-10.95	
14		10.95-11.65	

ASTER Bands



# Calculation of Surface Albedo from ASTER data

computed for each pixel of the ASTER image using the relation of Liang (2000):

$$\alpha = 0.484 \rho_1 + 0.335 \rho_3 - 0.324 \rho_5 + 0.551 \rho_6 + 0.305 \rho_8 - 0.367 \rho_9 - 0.0015$$

$\rho_i$  = surface reflectance in the ASTER band  $i$

### Processing steps:

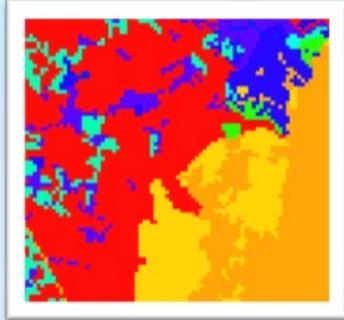
- Albedo was computed with a ground resolution of 30 m;
- The average value for each land cover class was calculated;
- The obtained values were used in place of the values contained in the parameterization scheme.

# Results

## Surface Albedo from ASTER data

Updated CORINE Classification

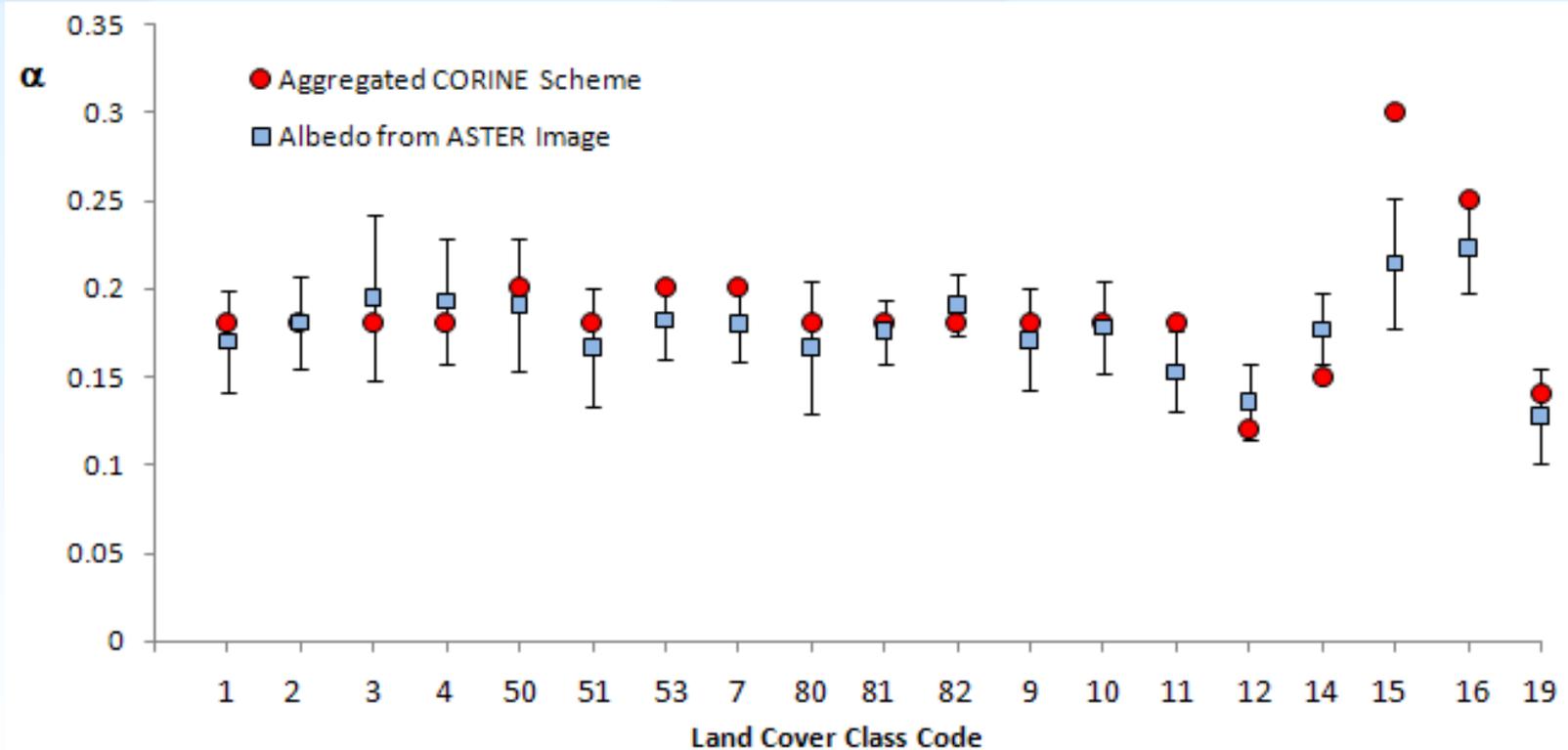
ASTER Albedo Image



Pixels  
belonging  
to class  $i$



Albedo Mean  
Value and  
Standard  
Deviation for  
Class  $i$



## Information retrieval 2

### Updating the CORINE dat set

### with ASTER retrieved land cover map

We developed a **set of rules and procedures**, with the goal of increasing the repeatability of the process, reducing errors and subjectivity.

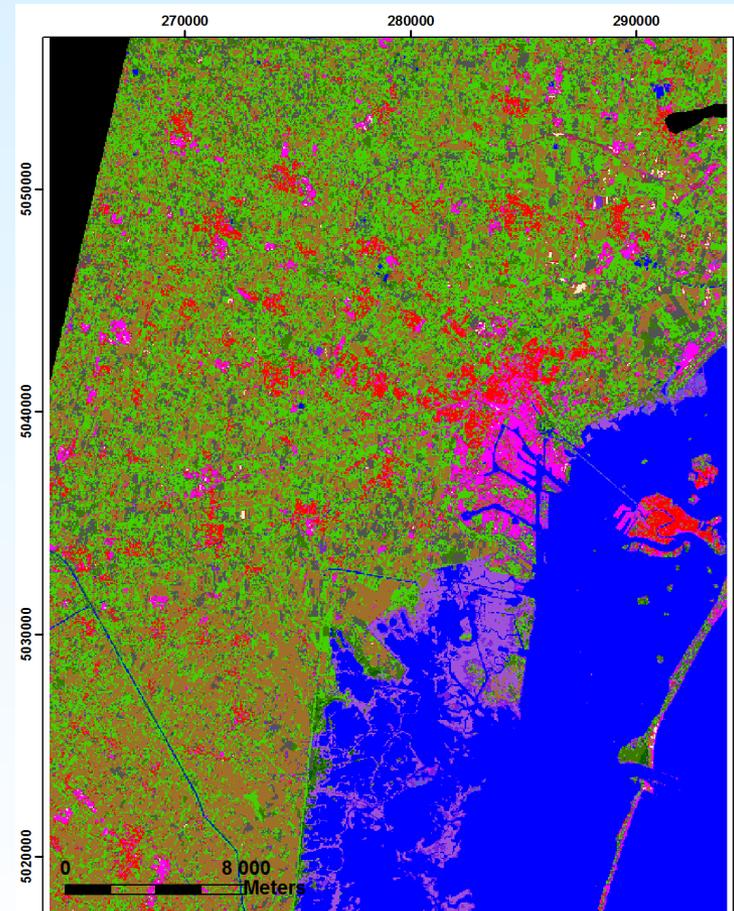
- **Standard legend of cover types** to be retrieved from the image. CRITERIA: classes easily separable with remote sensing and having significant differences in the bio-physical properties affecting surface-atmosphere interactions.

Legend for classification of Remote Sensing images
Urban
Industrial
Concrete
Fallow ground/Stubbles
Green vegetation (crops or grass) sparse/mix with dry vegetation
Green vegetation (crops or grass) quite homogeneous
Green vegetation (crops or grass) dense/homogeneous (mature)
Senescent dry vegetation dense/homogeneous (mature cereals)
Coniferous homogeneous canopy
Broadleaved homogeneous canopy
Mix coniferous-broadleaved homogeneous canopy
Broadleaved unhomogeneous wood
Woody crops
Coniferous unhomogeneous wood
Bare rock (fresh cut) (quarries)
Sand (beaches)
Exposed bare rock/dry bare ground with low humus content
Evergreen high and low bush/shrubs (Maquis)
Burnt areas
Lagoon - Emerged marsh vegetation
Lagoon - Wet bare ground
Lagoon - Wet vegetated ground
Water
Snow/Ice

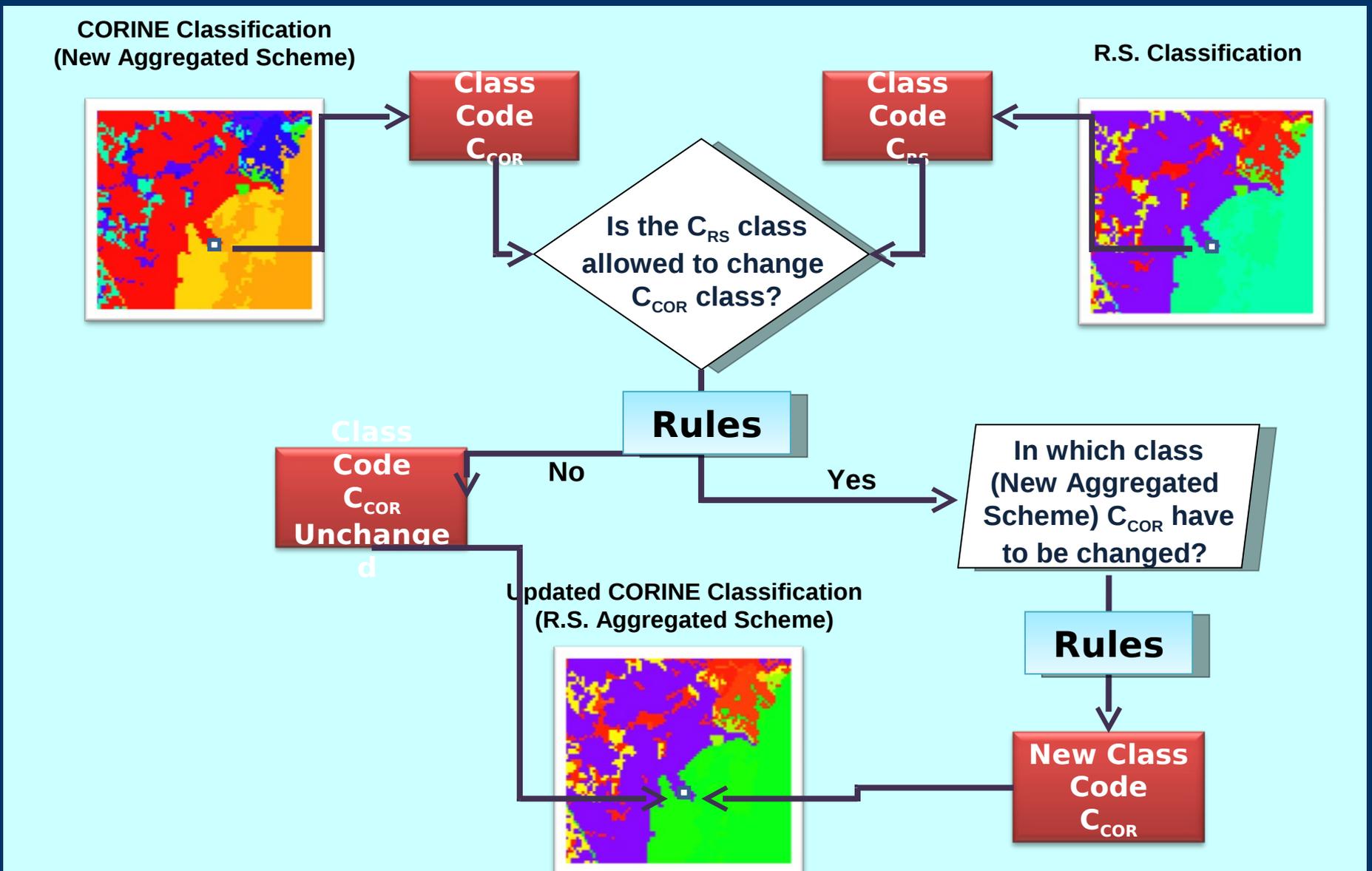
# Updating of the CORINE land cover map with ASTER information

We developed a **set of rules and procedures**, with the goal of increase the repeatability of the process, reducing errors and subjectivity.

- **Standard procedure of image classification**. CRITERIA: use of binary decisions wherever possible, to assign pixels to classes.

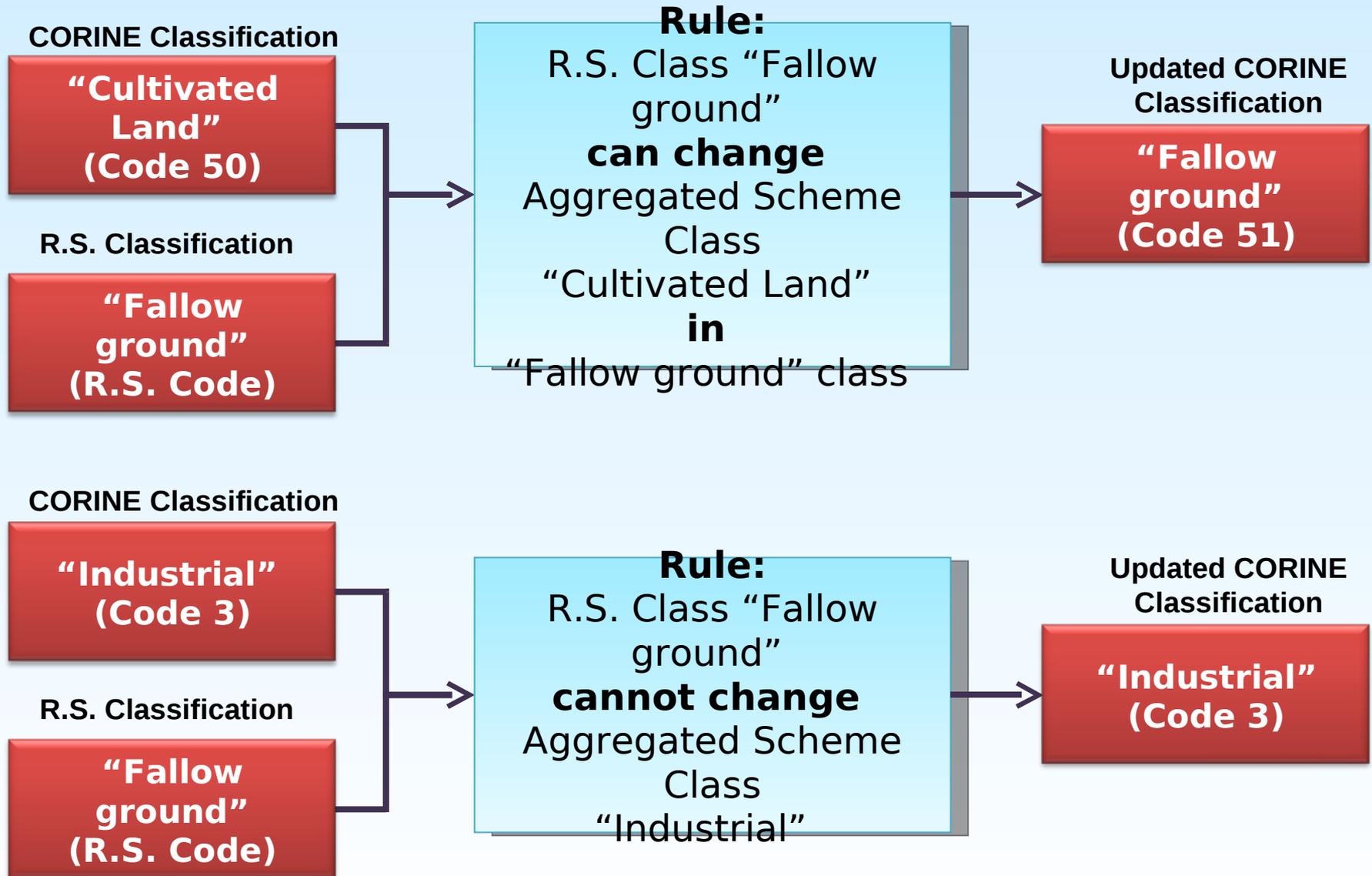


- Set of RULES** that control the way the remote sensing classification modifies the Aggregated CORINE Land Cover. CRITERIA: maximizing the input of significant information; minimizing the influence of possible R.S. classification errors.



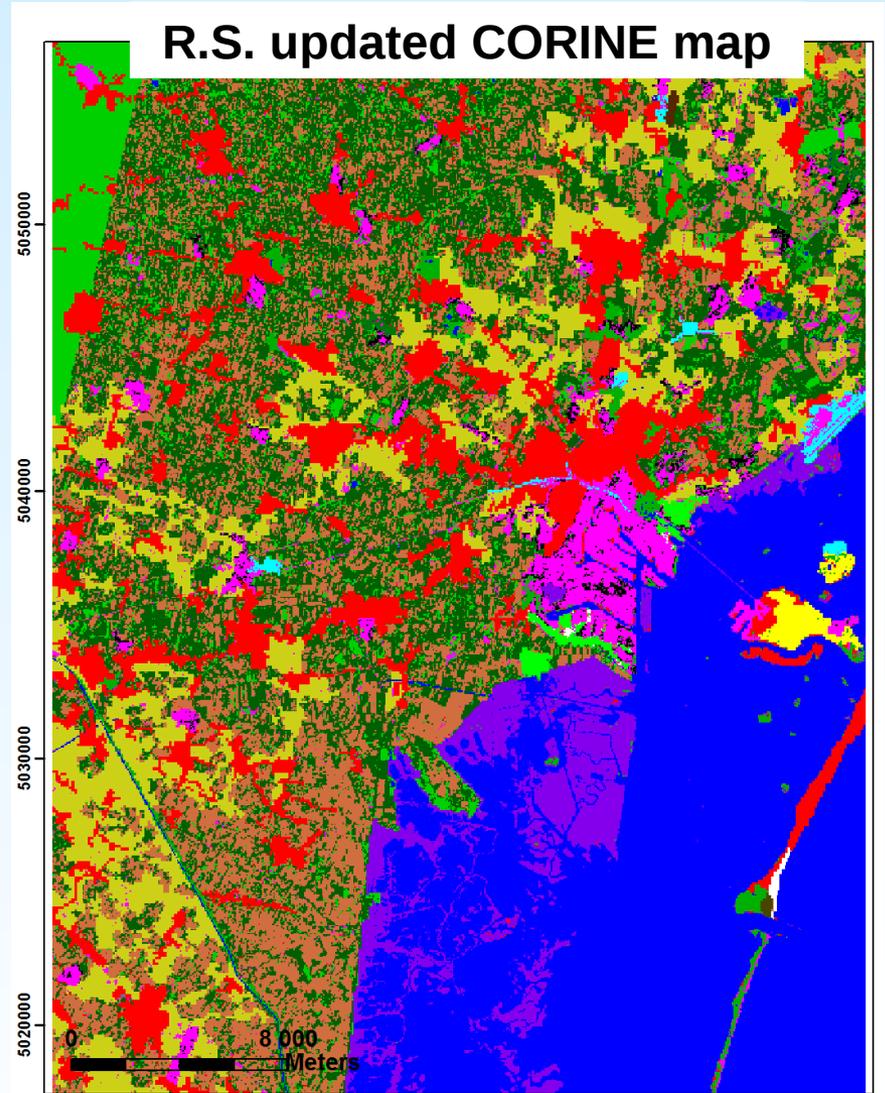
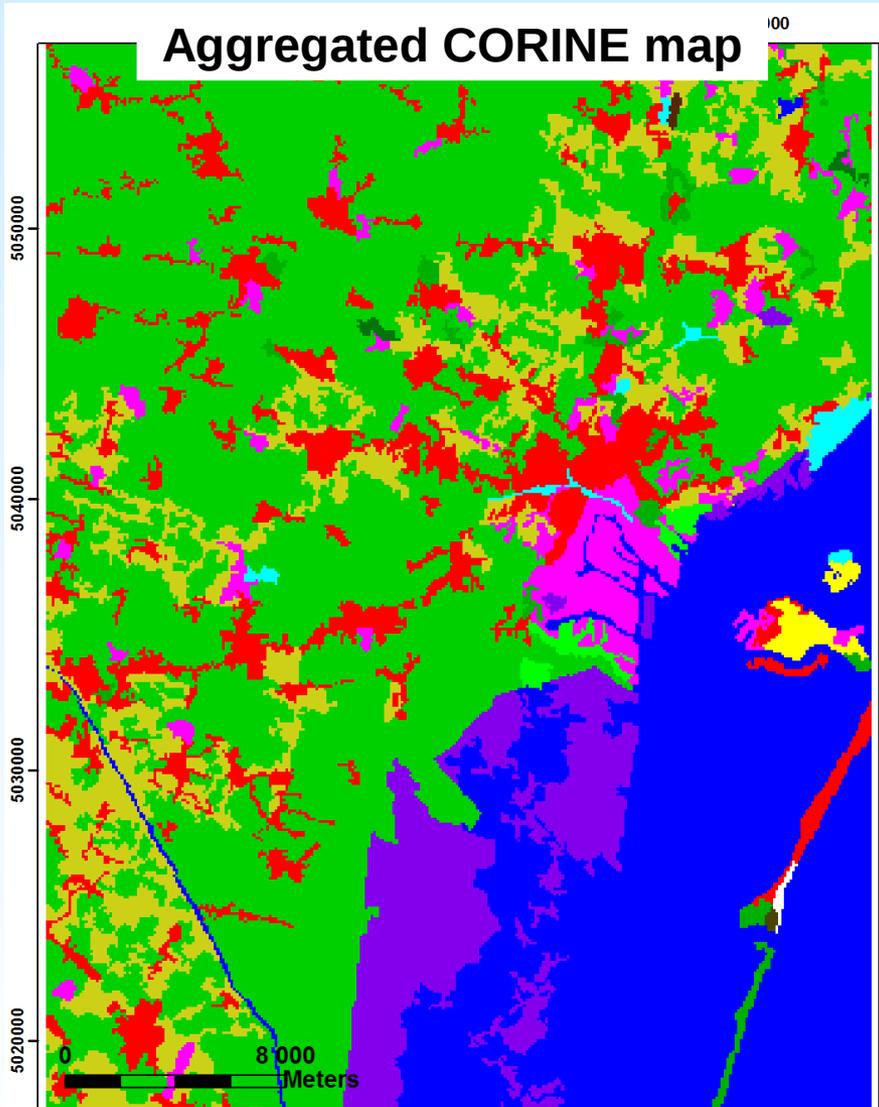
# Way the remote sensing classification modifies the Aggregated CORINE Land Cover.

## Examples of *RULES*



# Results

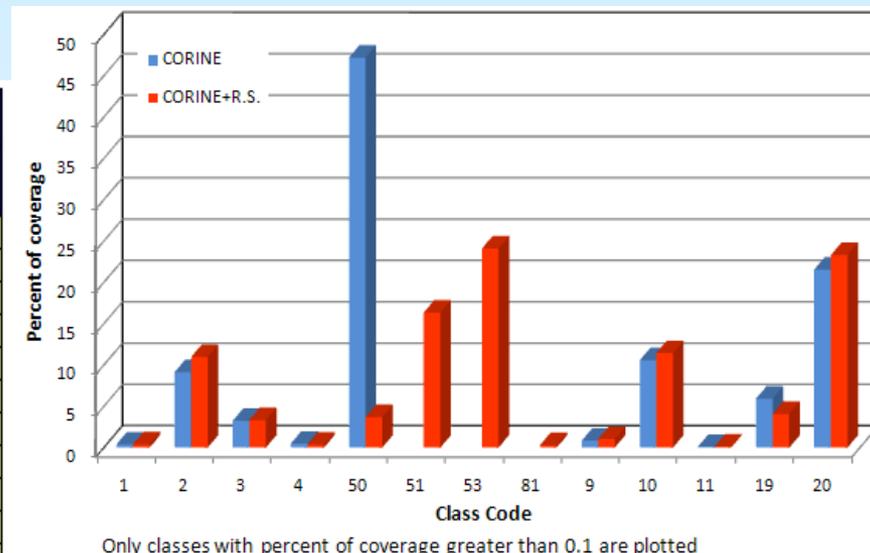
## Land Cover changes induced by Remote Sensing



# Results

## Land Cover changes induced by Remote Sensing

CODE	New Aggregated CORINE Scheme - R.S.	Percent of Coverage	
		CORINE	CORINE + R.S.
1	Continuous urban fabric	0.4	0.4
2	Discontinuous urban fabric	9.1	11.0
3	Industrial	3.2	3.3
4	Open, mostly artificial surfaces	0.5	0.4
50	Cultivated land - generic	47.2	3.7
51	Cultivated land: Fallow ground/Stubbles	-	16.3
52	Cultivated land: Very young crops	-	0.0
53	Cultivated land: Low mature green crops	-	24.1
54	Cultivated land: High senescent crops	-	0.0
6	Rice fields	0.0	0.0
7	Permanent/woody crops	0.0	0.0
80	Pastures - generic	0.0	0.0
81	Pastures: Low-sparse grass	-	0.3
82	Pastures: Thick/homogeneous grass	-	0.0
9	Crops/wood mosaic	0.8	1.0
10	Scattered settlement (farms, villages, trees, ...)	10.6	11.4
11	Broadleaved forest	0.1	0.1
12	Coniferous forest	0.0	0.0
13	Mixed forest	0.0	0.0
14	Deciduous or evergreen Bush/Shrub vegetation	0.4	0.4
15	Beaches, dunes and sand plains	0.0	0.0
16	Bare rock	0.0	0.0
17	Sparsely vegetated areas	0.0	0.0
18	Glaciers and perpetual snow	0.0	0.0
19	Inland and coastal wetlands	5.9	4.1
20	Water bodies	21.5	23.3



Percents of Coverage have been extracted from the land cover maps given as input to the dispersion model (250 m x 250 m grid)

# Results

## Effects of ASTER information on Dispersion results

Variable (units)	Run	Land, Day		Land, Night		Water, Day		Water, Night	
		Mean, $\sigma$	RMSD						
$z_0$ (cm)	R0	32.8, 30.8	17.3	32.8, 30.8	17.3	0.5, 2.3	6.5	0.5, 2.3	6.5
	R1	31.5, 34.4		31.5, 34.4		0.6, 6.0		0.6, 6.0	
$u^*$ (m s <sup>-1</sup> )	R0	0.50, 0.21	0.10	0.25, 0.16	0.07	0.43, 0.27	0.19	0.24, 0.14	0.13
	R1	0.49, 0.21		0.25, 0.17		0.49, 0.25		0.29, 0.18	
$w^*$ (m s <sup>-1</sup> )	R0	1.07, 0.63	0.12	0.03, 0.11	0.01	0.86, 0.39	0.34	0.72, 0.38	0.42
	R1	1.11, 0.65		0.03, 0.11		0.99, 0.36		0.89, 0.39	
$L$ (m)	R0	-206, 357	125	45, 173	73	-278, 342	267	-30, 111	132
	R1	-190, 343		46, 177		-304, 321		-48, 138	
$h$ (m)	R0	1163, 681	111	162, 130	50	676, 398	256	346, 208	180
	R1	1202, 713		163, 136		772, 370		435, 244	

$$RMSD = \sqrt{\text{mean} \left( f_{R_1} - f_{R_0} \right)^2}$$

$f$ : model output

R0 = reference run

R1 = run using remote sensing

$z_0$ : roughness length

$u^*$ : friction velocity

$w^*$ : convective scale velocity

$L$ : Monin-Obukhov length

$h$ : mixing height

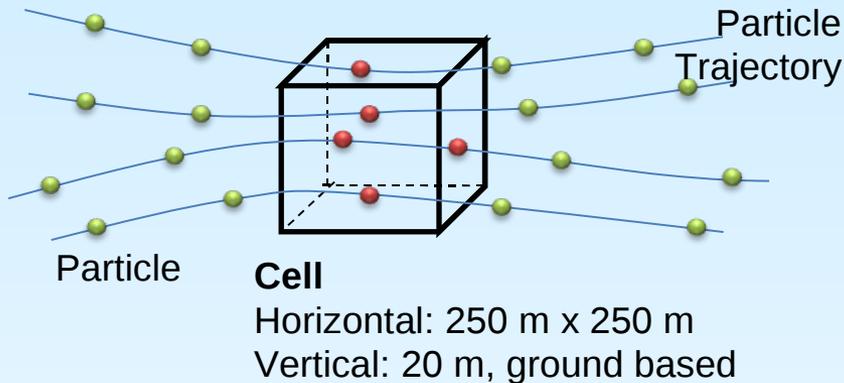
R.S. changed the spatial distribution of the outputs.

Their statistic distribution is little affected.

Changes are more important over water than over land,  
and during day than during night.

# Results

## Effects of ASTER information on Particle distribution



$n_p$  = Number of particles that transited through a given surface cell

$$\max(n_p) = 16$$

$$\overline{n_p} = 1.2$$

$$\sigma_{n_p} = 1.8$$

mean and standard deviations obtained from the two runs differ by less than  $10^{-2}$

Number of surface cells ( $n$ ) crossed by a n. of particles greater than a given threshold  $n_t$

Run	$n_t = 5$	$n_t = 6$	$n_t = 7$	$n_t = 8$	$n_t = 9$	$n_t = 10$
R0	$n = 1202$	$n = 710$	$n = 402$	$n = 213$	$n = 123$	$n = 74$
R1	$n = 1166$	$n = 659$	$n = 349$	$n = 185$	$n = 97$	$n = 55$
$100 \cdot (n_{R1} - n_{R0}) / n_{R0}$	-3%	-6%	-13%	-13%	-21%	-26%

R0 = reference run

R1 = run using remote sensing

The reference run produces more cells with high number of particles than run R1

→ the Run using R.S. produces more dispersion

# Conclusions



Results for a single case study!!

## Updating the land use map with ASTER classification

- **Very small** differences on the **mean values** of the outputs of the model;
- **Significant changes** in their **distribution** over the domain.
- Causes of these differences should be further investigated.

## Using ASTER albedo in the model

- Satellite albedo values are **very close** to the values contained in the model **internal tables**;  Remote sensing images can be used as an alternative way to obtain reliable values of this parameter.

## Further work

- Application to different case studies, to confirm the present first results;
- Attempt to extract other surface parameters needed by the models.