

Rural urban interface scenarios in Portugal based on land cover changes



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Motivation

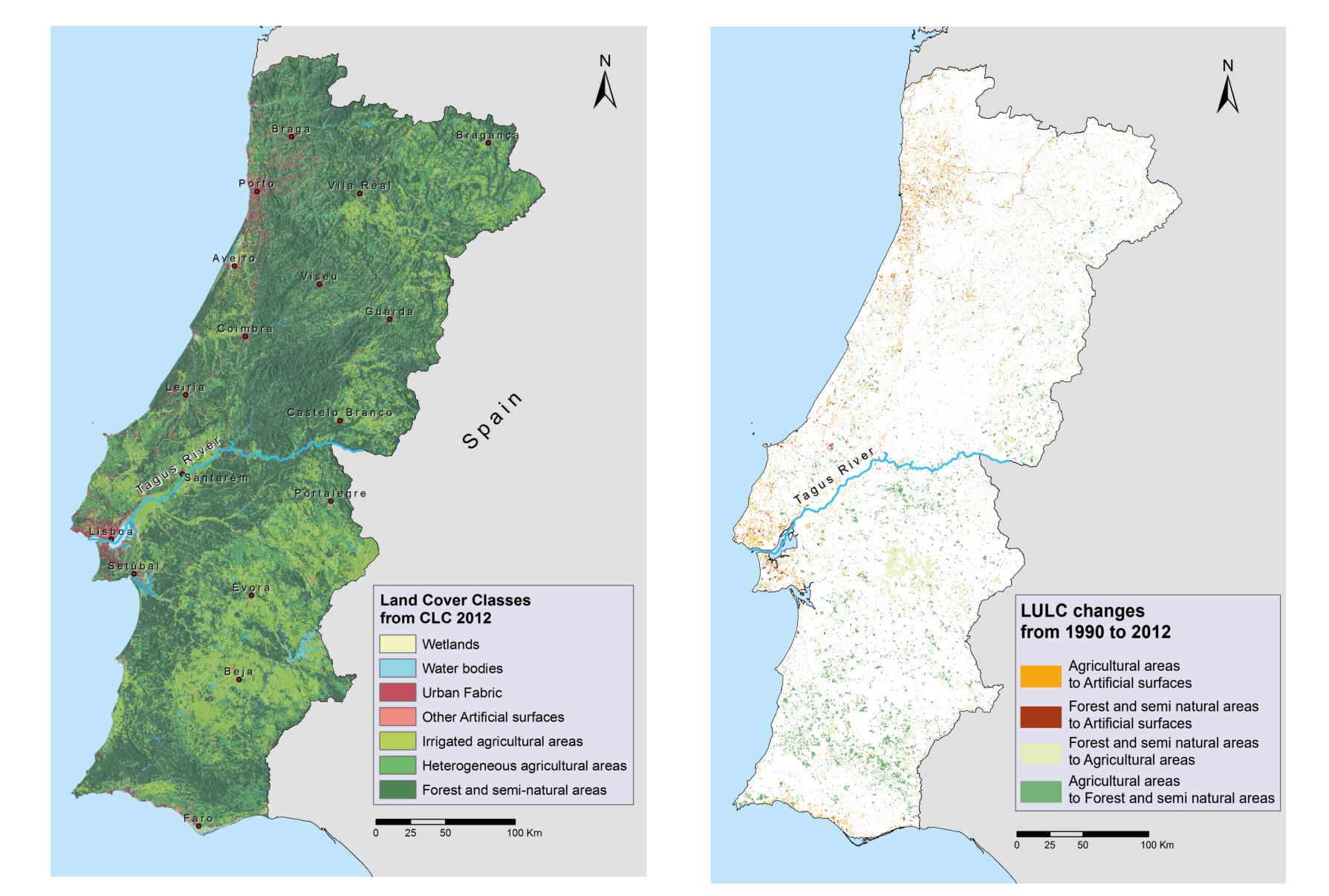
The Rural Urban Interface (RUI) define the area where humans and their development intermix with wildland fuels, including forest and rural areas. Here human-caused wildfires are more likely to occur and represent a main hazard for people, houses and infrastructures.

Land use/land cover changes (LULCC) highly affect the spatio-temporal evolution of the RUI.

RUI mapping is generally based on measurements the distance among specific land covers (i.e. urban area and forest/rural vegetation), but this methodology suffers from the definition of fixed parameters. To avoid this arbitrariness, a new procedure based on Multilayer Perceptron (MLP) and Fuzzy Set Theory is introduced in this study.

The methodology is applied to the case study of Portugal.

Study area



Data Source: Corine Land Cover (CLC) inventory (Copernicus Programme, http://land.copernicus.eu)

Mainland Portugal has a surface of 89,000 km² with an altitude range from sea level to about 2,000 m and a temperate climate.

The country is splitted by Tagus River into the northern half, where forest is predominant, and the southern half, where prevail agricultural lands and scrub vegetation. Main urban areas are mostly located close to the coast.

The northern half (with 25,322 fires registered in the period 1990-2013) is much more affected by wildfires than the southern (with olny 1,951 event in the same period).

LULC information came from the CORINE Land Cover inventory (CLC). Driving variables were also used to calibrate the simulation model (e.g. DEM, census data, road network, soil properties).

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General flowchart of the methodology

Three maps, cover Land Cover TO and T1 Static Driving Variables Dynamic Driving Variables corresponding to $T_0 = CLC_{1990}, T_1 =$ CLC_{2000} and $T_2 = CLC_{2012}$, were used Multilayer Perceptron for Transition Transition Potential Matrix T1-T2 to calibrate and validate the model. Modelling CLC1990 and CLC2000 acted as input to Land Cover Prediction T2 Urban Growth probability map (T2) Rural Growth probability map (T2) prenict a land cover maps at the time T₂, adopting **MLP** neural network algorithm for the **transition** ______ Fuzzy Overlay and potential modelling. Land Cover T2 Possibility Map Creation was than The resulting map Rural-Urban Interface T2 compared with the real CLC₂₀₁₂. Once Kappa Statistics the model validated, via **Kappa** Area Under the Curve statistics, MLP the same hyper-parameters were applied to Is the simulation simulation simulate the land use map at the time correct? T₃, by means of the CLC₂₀₀₀ and CLC2102. At this stage, Markov Chain procedure was applied to allocate the Multilayer Perceptron for Markov Chain for Change **Transition Potential** transition. Allocation Matrix T2-Finally, the **future scenario** allowing Land Cover Prediction T3 Rural Growth to predict the land cover for the year 2030 (i.e. LULC hard prediction), **Fuzzy Membership Function** Urban Growth probability map (T3) Rural Growth probability map (T3) Fuzzy Overlay (T3) provided new boundaries for the future RUI map. Fuzzy set theory was applied to expressing the produce maps Rural-Urban Interface 2012 **Transition Potential** All-to-Heterogeneous All-to-Forests probability of each pixel to transit Logical OR toward rural coverages or urban areas in 2030 (i.e. soft prediction). Transition Potentia All-to-Urban ransition Potentia All-to-Rural These maps were then overlapped through joint membership fuzzy mean "urban" value per training set Extract pixel values in Training Set: 70% Testing Set: 30% mean "rural" value per training set each point functions, resulting in map expressing the **possibility** of an area to belong to the **RUI in 2030** (i.e. Fuzzy Or overlay the RUI possibilty map). Fuzzy near membership function Fuzzy Sum overlay fuzzifucation validate the Fuzzy large membership function AUC and best model procedure, the RUI bounderied were Fuzzy Gamma selection used as benchmark vs the simulated RUI2012 and the AUC was computed. idation and model select

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Results: soft and hard prediction

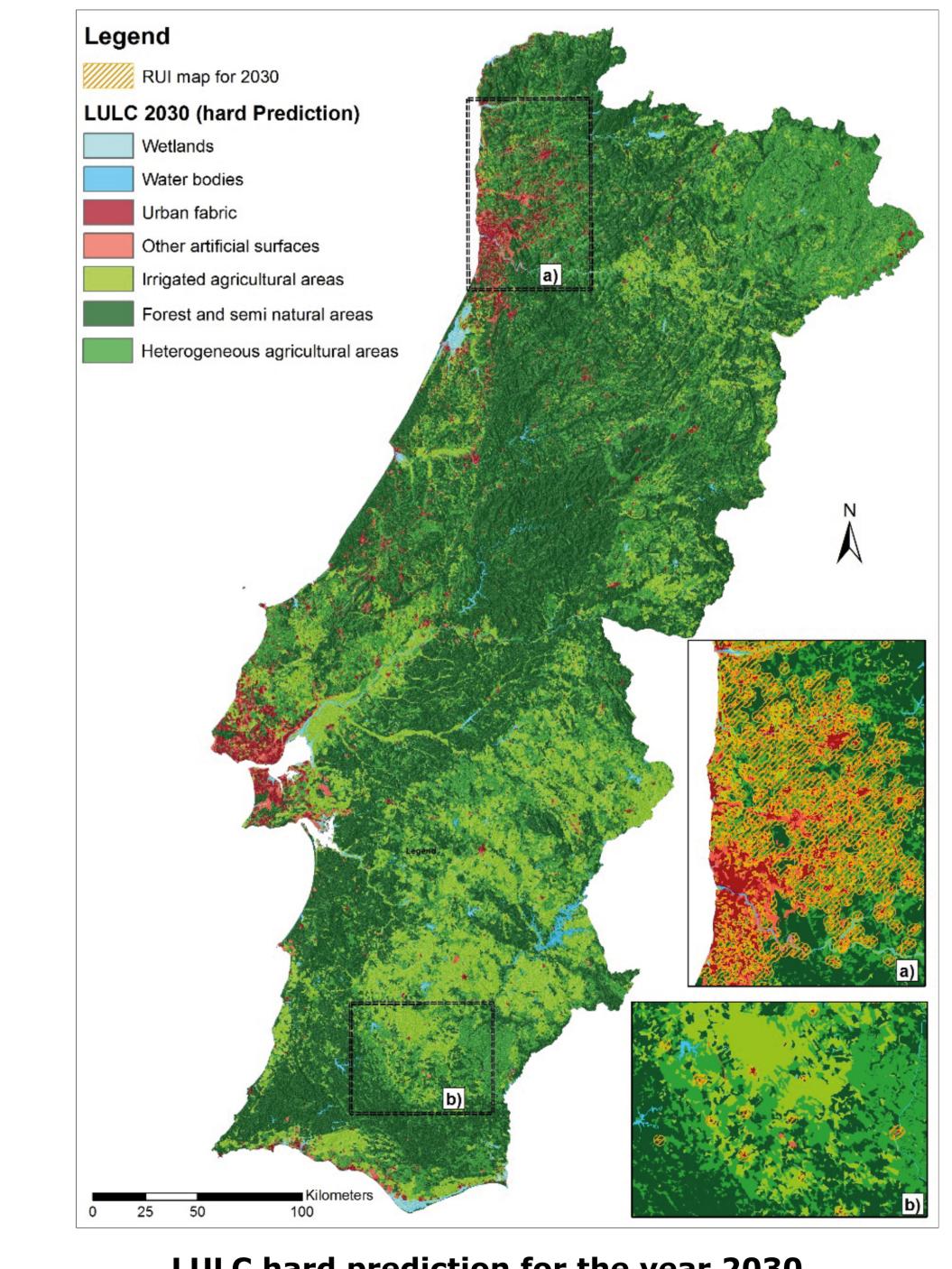
The proposed methodology resulted into two land cover scenarios for 2030, allowing to predict the future RUI: • Hard Prediction ----- future RUI boundaries

- Soft Prediction ----- possibility fuzzy RUI map

The difference between the two is that the soft prediction yields the entire set of simulated transitions, while the hard prediction yields only one specific transition, selected through a multi-objective land competition model.

The areas where RUI has the higher possibility to extend in 2030 were find in the fringes of urban areas, driven by the peri-urban growing. Specifically, in the area enclosed by Braga, Porto, Aveiro and Vila Real in the North-West, with a spatial contiguity along the coast up to the city of Lisbon. In the South, highly predisposed area is along the Faro Region, probably due to the expansion of urbanism in this costal-touristic place. Eastern-Northern interior mountainous lands present scattered hotspots for RUI, caused by the abandonment of agricultural lands and the consequent forest spreading.

classes (i.e. rural and urban) identified on the prospective CLC2030. advantage of not relying on expert knowledge inputs.



LULC hard prediction for the year 2030 Zoomed areas represent the RUI boundaries (considering a buffer width of 1000 m) in a highly densely populates places around Porto (a) and in scattered rural-residential areas (b).



The resulting **possibility fuzzy RUI map** represents the tendency towards the overlapping areas between relevant

This innovative methodology, although quite complex, is extremely robust, account for uncertainty and has the

