

VU Research Portal

Mediterranean land systems: Representing diversity and intensity of complex land systems in a dynamic region

Malek, Ziga; Verburg, P.H.

published in

Landscape and Urban Planning
2017

DOI (link to publisher)

[10.1016/j.landurbplan.2017.05.012](https://doi.org/10.1016/j.landurbplan.2017.05.012)

document version

Peer reviewed version

document license

CC BY-NC-ND

[Link to publication in VU Research Portal](#)

citation for published version (APA)

Malek, Z., & Verburg, P. H. (2017). Mediterranean land systems: Representing diversity and intensity of complex land systems in a dynamic region. *Landscape and Urban Planning*, 165, 102-116.
<https://doi.org/10.1016/j.landurbplan.2017.05.012>

General rights

Copyright and moral rights for the publications made accessible in the public portal are retained by the authors and/or other copyright owners and it is a condition of accessing publications that users recognise and abide by the legal requirements associated with these rights.

- Users may download and print one copy of any publication from the public portal for the purpose of private study or research.
- You may not further distribute the material or use it for any profit-making activity or commercial gain
- You may freely distribute the URL identifying the publication in the public portal ?

Take down policy

If you believe that this document breaches copyright please contact us providing details, and we will remove access to the work immediately and investigate your claim.

E-mail address:

vuresearchportal.ub@vu.nl

Mediterranean land systems: representing diversity and intensity of complex land systems in a dynamic region

Authors: Žiga Malek, Peter Verburg

Affiliation: Environmental Geography Group, Department of Earth Sciences, Vrije Universiteit Amsterdam, De Boelelaan 1087, 1081 HV Amsterdam, the Netherlands

Corresponding author:

Žiga Malek

Tel: +31 20 5988710

Fax: +31 20 6462457

z.malek@vu.nl

Co-author contact:

peter.verburg@vu.nl

1 **Abstract**

2 In the Mediterranean region, land systems have been shaped gradually through centuries. They
3 provide services to a large and growing population in a region that is among the most vulnerable
4 to future global change. The spatial extent and distribution of Mediterranean land systems is,
5 however, unknown. In this paper, we present a new, expert-based classification of Mediterranean
6 land systems, representing landscapes as integrated social-ecological systems. We combined data
7 on land cover, management intensity and livestock available on the European and global scale in
8 a geographic information system based approach. We put special emphasis on agro-silvo-
9 pastoral mosaic systems: multifunctional Mediterranean landscapes hosting different human
10 activities that are not represented in common land cover maps. By analyzing location conditions
11 of the identified land systems, we demonstrated the significance of both bio-physical
12 (precipitation, soil) and socio-economic (population density, market influence) factors driving
13 the occurrence of these systems. Agro-silvo-pastoral mosaic systems were estimated to cover
14 23.3% of the Mediterranean ecoregion and exhibited to a certain extent similar characteristics as
15 forest and cropland systems. A reanalysis using data that are available with global coverage
16 indicated that the choice of datasets leads to significant uncertainties in the extent and spatial
17 pattern of these systems. The resulting land systems typology can be used to prioritize and
18 protect landscapes of high cultural and environmental significance.

19 **1. Introduction**

20 In light of recent socio-economic developments and anticipated climate change impacts in the
21 Mediterranean region, there is an urgent need for investigating the capacity of the region to
22 sustain a variety of ecosystem services for a growing population. On one side, the European part
23 of the region is home to high-input intensive agricultural systems significant for regional food
24 production. On the other, the Middle Eastern and North African part is among the regions with
25 the highest population growth, and dependency on food imports - with over half of the
26 population relying on food produced elsewhere (Wright & Cafiero, 2011). The region is
27 extremely vulnerable to fluctuations in food supply and prices, and expected climate change
28 coupled with demographic growth could contribute to further regional instability and conflicts
29 (Evans, 2008; Sowers *et al.*, 2010). Potential shocks to the society and economy have also been
30 observed in the European part. The Greek financial crisis reportedly influenced the supply of
31 agricultural products (Pfeiffer & Koutantou, 2015), impacting on land-use and environment.

32 In order to target policies to prioritize areas for agriculture, landscape conservation and
33 biodiversity protection in the Mediterranean region, the characteristics and distribution of land
34 systems need to be identified (Agnoletti, 2014). This is particularly valid for agro-silvo-pastoral
35 mosaic systems where human influence and ecological conditions are intricately linked.
36 Characteristics of such traditional landscapes are disregarded if represented by a single,
37 dominant land cover type as is common in most current datasets (Turner *et al.*, 2007; Verburg *et*
38 *al.*, 2011a). Moreover, when analyzing changes to these systems, land-use intensity is an
39 important component besides changes in land cover, and has a significant environmental impact
40 (Ellis & Ramankutty, 2008). Existing land cover and land systems mapping approaches are
41 misrepresenting the extent or diversity of agro-silvo-pastoral mosaics (Zomer *et al.*, 2009) and

42 often fail to integrate differences in land-use intensity. Although global and continental attempts
43 to map land systems in the Mediterranean region were made, they focused on generalized
44 cropland and grazing systems (van Asselen & Verburg, 2012; Dixon *et al.*, 2001; FAO, 2011),
45 ignoring the specific mosaics unique to this region.

46 As a result of its environmental conditions, extremely long land-use history, and cultural
47 diversity, the Mediterranean region is characterized by a wide variety of land systems that are not
48 easily mapped. A good example is the dehesa/montado system, present in Spain and Portugal,
49 which is highly valuable in the cultural heritage context (Meeus, 1995). In this system different
50 activities, such as gathering of forest products, livestock grazing and cereal cultivation occur
51 simultaneously (Joffre *et al.*, 1999). Using remote sensing imagery, we can receive information
52 on the tree density of these systems, but not on the extent of grazing or crop cultivation below
53 the trees (Plieninger & Schaar, 2008). Attempts to map these multifunctional systems have been
54 made. In the European CORINE land cover data, they are represented as “Agroforestry areas”,
55 however substantial areas are also defined as other classes (Bunce *et al.*, 2008; EEA, 2015a).

56 In the Mediterranean region, landscapes are subject to two contrasting processes of change:
57 abandonment of rural, mountainous and less developed areas on one side, and intensification and
58 increasing human influence on the other (García-Llorente *et al.*, 2012; Nieto-Romero *et al.*,
59 2014). Soil degradation and water shortages are the main environmental problems in the region,
60 as a consequence of land management and complex biophysical and climatic conditions
61 (Almagro *et al.*, 2013; Guerra *et al.*, 2015). Furthermore, projected climate and socio-economic
62 changes suggest that Mediterranean ecosystems are amongst the most vulnerable to future global
63 change (Schröter *et al.*, 2005). Traditional agro-silvo-pastoral mosaic systems are particularly
64 under pressure, threatening the provision of numerous ecosystem services and biodiversity in

65 general (Zamora *et al.*, 2007). A significant number of plant and animals species, a lot of them
66 endemic, are related to extensive management practices and these traditional landscapes. This is
67 why the Mediterranean region was identified as one of the Global Biodiversity Hotspots
68 (Cuttelod *et al.*, 2009).

69 In this paper we develop a spatial representation of Mediterranean land systems by integrating
70 information on land management as an inseparable part of these landscapes. By investigating the
71 location factors behind these land systems, we identify how different socio-economic and
72 biophysical factors determine their distribution. At the same time, this study addresses the
73 challenges of data and knowledge differences between different parts of the Mediterranean
74 region. Finally, we evaluate the performance of our classification, by comparing it to existing
75 studies in the region, and by analyzing the uncertainty related to available data.

76 **2. Materials and methods**

77 **2.1 Study area**

78
79 We defined the spatial extent of the Mediterranean region by focusing on areas surrounding the
80 Mediterranean Sea that share similar climatic and other biophysical characteristics. We chose the
81 spatial extent of the Mediterranean ecoregion (Fig. 1), as it describes the approximate extent of
82 representative Mediterranean natural communities (Olson *et al.*, 2001). We included the Nile
83 Delta and similar ecoregions within the Mediterranean ecoregion, such as the Apennine
84 deciduous montane forests in central Italy. The total study area covers 2.3 million km² in 27
85 countries. Around 400 million people live within the ecoregion boundaries, and yearly 250
86 million tourists visit the area (31% of all international tourists), making it among the regions with
87 highest human influence (Cuttelod *et al.*, 2009). The region is characterized by the

88 Mediterranean climate with dry summers and mild winters, when most precipitation takes place.
89 The southern part of the region is predominantly arid and semi-arid, whereas the northern part is
90 semi-arid to dry humid (Zomer *et al.*, 2008). Although the mean annual precipitation of the
91 whole area is around 500 mm, a quarter of the area has below 300 mm of rainfall. This limits
92 rainfed agriculture, particularly in the Middle East and North Africa part of the region.

93 **2.2 Classification overview**

94
95 We classified combinations of land cover, livestock density, irrigation extent and different
96 intensity proxies (Table 1) using a Geographic Information System (GIS) based approach. By
97 combining land cover with data on land management, we considered the anthropogenic aspects
98 of Mediterranean land systems. This is necessary, as the management of a specific location
99 depends on local combinations of socio-economic and biophysical conditions (Lambin *et al.*,
100 2001). Mediterranean land system classes were defined a-priori based on the common types
101 distinguished in the literature.

102 We operated on a 2 km spatial resolution. Although a 2 km spatial resolution is arbitrary this
103 would hold for any chosen resolution that aims to capture human-environment interactions. The
104 choice of spatial resolution was based on: 1). The continental extent of the Mediterranean region
105 and the spatial detail of available data. Although some of the data were available on a very high
106 resolution (e.g. 25 m tree cover), most of it was available on a 1 km resolution (Table 1); 2).
107 Land systems were defined by the set of activities at the farm or landscape level and not at the
108 level of individual landscape components (Verburg *et al.*, 2002), given the relatively small scale
109 and high spatial variation within landscapes a 2 km spatial resolution was judged to be optimal
110 for capturing variation in land systems; and 3). We aimed to represent global patterns of

111 Mediterranean land systems on a resolution able to capture the spatial variability of human-
112 environment interactions in heterogeneous landscape mosaics (van Delden *et al.*, 2011; Pickett &
113 Cadenasso, 1995).

114 **2.3 Data**

115
116 More data and data with higher thematic and spatial resolution were available for the European
117 part of the region (Fig. 1). In contrast to studies that only use data that are consistently available
118 across an entire study area, we used the best data available for different parts of the region.
119 However, we restricted ourselves to data that covered multiple countries. National data were
120 used to train the classification (e.g. by looking at the dehesa/montado extent). The following
121 criteria were used when choosing the data: 1). Highest spatial resolution; 2). Data were as recent
122 as possible; 3). Data underwent validation; 4). The data were not generated by downscaling
123 based on population density. This way we could ensure independence of the data and later
124 analyze how the occurrence of land systems relates to population distributions. All input maps
125 were resampled to a resolution of 2 x 2 km in an Lambert equal area projection.

126 For land cover variables, we used tree cover (Hansen *et al.*, 2013), soil sealing data for Europe
127 (EEA, 2015b), built up areas extent for the remaining part of the region (Jun *et al.*, 2014),
128 cropland extent (Fritz *et al.*, 2015) and the extent of bare areas (Latham *et al.*, 2014). For
129 identifying the extent of areas with permanent crops, we used the CAPRI-dynaspat data for the
130 European Union part of the region (Britz & Witzke, 2014), the CORINE land cover permanent
131 crops extent for the rest of Europe and Turkey (EEA, 2015a), and the SPAM data for the MENA
132 region (You *et al.*, 2014).

133 Livestock distribution was obtained from the Gridded Livestock of the World v2.0 (Robinson *et*
134 *al.*, 2014). We combined the numbers of bovines, goats and sheep. Livestock distribution was
135 used to identify rangelands and grazing mosaic systems, and to define the intensity of grazing
136 based on an existing grazing systems classification (Dixon *et al.*, 2001; FAO, 2011). We did not
137 consider the distribution of pigs. Pigs are being grazed on a large extent in the dehesas/montados
138 of the Iberian peninsula, where they are associated with traditional products such as the “jamón”.
139 Pigs in other parts of the Mediterranean are mostly attributed to landless livestock management
140 patterns. Based on the data these two different systems could not be distinguished.

141 Irrigation plays a significant role in the Mediterranean region, where agriculture is constrained
142 by water availability (Almeida *et al.*, 2013). Although irrigation cannot be related to agricultural
143 intensity, irrigated systems have specific demands regarding water and energy (Fader *et al.*,
144 2016). To map irrigated systems, we used the data on areas equipped for irrigation from the
145 Global Map of Irrigation Areas (Siebert *et al.*, 2005, 2013).

146 We used different indicators and proxies to characterize the intensity of land management, as
147 data on this spatial scale is scarce. We used the European agricultural intensity map to identify
148 areas with intensive rainfed cropland for the European Union part of the Mediterranean region
149 (Temme & Verburg, 2011). For the remaining area, we used the global field size map, where we
150 defined the areas with the largest field size class as intensive (Fritz *et al.*, 2015). While it is not
151 possible to directly translate field size to intensity, field sizes can indicate the degree of
152 investment, mechanization and labor intensity of agriculture (Kuemmerle *et al.*, 2013; Rodríguez
153 & Wiegand, 2009). In addition, areas within the 10th percentile of crop yields in the non-EU
154 Mediterranean region were identified as intensive. We focused on the most significant crops in
155 the Mediterranean region: wheat and other cereals, together with vegetables for annual crops;

156 and tropical and temperate fruits (among them grapes), together with olives for permanent crops
157 (Daccache *et al.*, 2014).

158 For forest management intensity, we used the European forest management map with defined
159 areas of high forest harvesting intensity (Hengeveld *et al.*, 2012). We identified planted forests
160 by looking at areas with a high share of plantation species using the European tree species map
161 (Brus *et al.*, 2012; Verkerk *et al.*, 2015). For the non-European part of the Mediterranean region,
162 no such data is available. Therefore we used the forest losses and gains data between 2000 and
163 2014 to identify areas with high intensity of forest management, defined by the cycles of felling
164 and replanting. If the landscape, defined by the 2 km spatial resolution, experienced both high
165 losses and high gains in the observed time, we assumed it being a high intensity forest. If a
166 significant increase of forests occurred in the observed time, we defined it as a planted forest. We
167 assumed it is unlikely, that in a semi-arid environment vast areas would be reforested naturally in
168 such a short time.

169 **2.4 Expert-based classification**

170
171 We used an expert-based hierarchical classification procedure (Fig. 2, Supplement A).
172 Classification rules were defined as conditional thresholds based on literature on Mediterranean
173 farming, grazing, agro-silvo-pastoral and forest systems (full list of literature considered in
174 Supplement B). This way, our classification followed common understanding of the
175 characteristics of Mediterranean land systems. Expert-based hierarchical classification
176 procedures have been used to identify land and farming systems in numerous cases (Dixon *et al.*,
177 2001; van de Steeg *et al.*, 2010). We follow a similar classification procedure as van Asselen and
178 Verburg (2012) and the LADA project (FAO, 2011). However, none of these approaches dealt

179 with complex mosaic systems specific for the Mediterranean. Compared to statistical clustering
180 classification (Ellis & Ramankutty, 2008; Letourneau *et al.*, 2012; Václavík *et al.*, 2013), expert
181 based classification is less sensitive to the selected distance metric and criteria for determining
182 the order of clustering (van Asselen & Verburg, 2012). A detailed comparison between expert-
183 based and statistically derived typologies for landscapes is provided by van der Zanden *et al.*
184 (2016). Our hierarchy was based on management intensity. Land systems were identified using
185 different intensity indicators, and systems with low intensities were defined as areas where these
186 indicators do not show a high intensity. More intensive systems overwrote less intensive ones,
187 when more than one system fulfilled the classification criteria.

188 First, we defined settlement systems as areas with a high percentage of built-up areas. On a 9 cell
189 neighborhood we performed focal statistics and subsequently applied a majority filter to the
190 European sealed soil and the global land cover 30 maps. By looking at the immediate
191 neighboring cells as well, we identified larger built-up landscapes and removed individual cells
192 with high shares of built-up areas. Other systems that were defined by the dominant land cover
193 were systems occurring on bare (desert) areas, and wetlands (Supplement A). If later in the
194 classification stage we identified a high intensity cropland system at the same location as a
195 wetland, it was overwritten. For example, the Guadalquivir river estuary is defined as a wetland,
196 however a large portion of it is cultivated. This way, we resolved inconsistencies between data
197 sets and differences in definition (the high intensity cropping system is still in a wetland area).
198 After this step we continued with the classification of cropland, forest, grazing systems and agro-
199 silvo-pastoral mosaics.

200 Cropland, Forest, grazing and agro-silvo-pastoral mosaic land systems were at first defined by
201 the cropland extent and tree cover. Cropland systems were associated with high cropland extent

202 and were further subdivided depending on their intensity, presence of irrigation and
203 combinations of crop type. Forest systems occur on areas with a high tree density, and were
204 subdivided based on their protection status and harvesting intensity. Grazing systems were
205 subdivided based on whether they occur in semiarid or arid areas or grasslands, and their
206 livestock density.

207 The remaining agro-silvo-pastoral mosaic systems represent multifunctional agroforestry
208 landscapes. We identified them by looking at the activities they host: cropland, livestock grazing,
209 woodlands. We classified them based on their tree cover (open or closed woodlands), cropland
210 extent, and livestock density.

211 **2.5 Analysis of location factors**

212
213 The observed distribution of land systems reflects the continuity of land management as a
214 response to socio-economic and biophysical conditions (Fuchs et al., 2013). We performed
215 binominal logistic regressions to investigate the role of these conditions. This way we could
216 calculate the probability of each location to host a specific land system, an approach often used
217 to explain existing land-use patterns (Letourneau *et al.*, 2012). Logistic regressions were
218 performed for all land systems separately using 20 variables (Table 2).

219 Biophysical variables describe the suitability for growing crops, encouraging or constraining
220 agricultural activities (Panagos *et al.*, 2013). We selected seven soil characteristics: sand, clay
221 and organic content, cation exchange capacity (CEC), pH, drainage and soil depth. We used the
222 soil characteristics valid for natural vegetation conditions to omit potential correlation between
223 e.g. forest cover and organic content (Stoorvogel *et al.*, 2016). We also tested the soil
224 characteristics of the current land cover situation. Temperature, precipitation, solar radiation and

225 potential evapotranspiration are climatic variables that limit growth of vegetation. Although
226 aridity limits the growth of vegetation, we had to omit the CGIAR aridity index map to avoid
227 multicollinearity (Zomer *et al.*, 2008) as it was highly correlated to precipitation (Pearson
228 correlation >0.9). Lastly, we studied how potential natural vegetation explains the natural
229 vegetation characteristics of land systems.

230 Socio-economic factors were represented by five variables. Population density and density of
231 rural population characterize the type of activities expected in an area, and the degree of human
232 impact (Neumann *et al.*, 2015). The market influence index specifies the capital available to
233 agricultural production, investing in its expansion or intensification (Verburg *et al.*, 2011b).
234 Accessibility to national and international markets is an indicator for the potential to market
235 goods provided by the land systems (Verburg *et al.*, 2011b). Finally, we investigated the role of
236 road infrastructure, by including the distance to roads.

237 The regression was performed on a balanced sample of 5% of all grid cells for each land system
238 (with a minimum sample of 1000 points - 500 for presence and 500 for absence). To reduce
239 spatial autocorrelation while retaining a sufficiently large sample size, we applied a minimum
240 distance of one cell (4 km) between the sample points. We performed a forward conditional
241 regression. We used the ROC (Receiver Operating Characteristic) as a measure for the goodness
242 of fit of our regression model. Multiple samples were taken to ensure robustness of the identified
243 relations. Only for very small land systems (e.g. planted forests) this was not possible. For none
244 of the land systems we found major differences between the results based on different samples.

245

246 **2.6 Classification performance and data uncertainty**

247
248 Assessing the performance of a land systems classification is a difficult task, and cannot be
249 performed using traditional approaches applied in remote sensing or spatial simulation. Any
250 classification system is as good as its potential use and the quality of the underlying data. For
251 example, validation using high resolution satellite images or land cover products could only be
252 used to identify the category of land system (forest, cropland systems), without validating the
253 intensity. For being a useful classification, identified land systems should correspond to common
254 descriptions of these systems and be related to land systems found in field studies. We performed
255 a documented expert based validation, where we gathered studies from the whole Mediterranean
256 region. We collected 190 studies on land management from peer reviewed papers, book chapters
257 and conference proceedings (Supplement B). The studies were selected based on the following
258 criteria: 1). The study clearly defined a land system characteristic, such as intensity or the mosaic
259 nature of the system (e.g. intensive tomato production, dehesa); 2). The study was associated to a
260 specific location (Mediterranean or nationwide studies were omitted); 3). It was based on an
261 actual system and not on experiment sites. We registered the locations of all studies, together
262 with the information of their land system characteristics (type, intensity, management). The
263 accuracy of the final land systems map was then assessed by comparing how well it represents
264 the documented land systems. Studies on urban areas (Mediterranean cities) were omitted, as
265 they completely correspond with the locations of cities and would falsely contribute to a higher
266 accuracy.

267 To analyze the uncertainty related to the differential quality of data, we applied the same
268 classification criteria using the data with the lowest quality but global coverage (Table 1). High
269 resolution data covering the European part of the region were thus not used. The two maps were

270 compared in terms of agreement or disagreement of quantity and location (Pontius & Santacruz,
271 2014).

272 **3. Results**

273 **3.1 Land systems**

274
275 The distribution of Mediterranean land systems is shown in Figs. 3, 4 and 5. Average values for
276 land systems in terms of bare, tree and cropland cover, and livestock density are presented in Fig.
277 6 and Supplement C.

278 *3.1.1 Bare and open grazing systems*

279 Bare and open grazing systems cover 22.6% of the Mediterranean region, mostly in North Africa
280 and the Middle East. They are divided into grazing systems in arid environments and grazing
281 systems in open rangelands. Arid systems are further subdivided into bare areas and deserts
282 without notable livestock presence, and extensive and intensive arid grazing. In some parts (e.g.
283 Syria), livestock density in deserts can reach over a 1000 heads of combined sheep, goats and
284 bovines per km². Open rangelands are subdivided into extensive and intensive, and occur
285 primarily in open landscapes of the Iberian peninsula, North Africa, Turkey and the Western
286 Balkans. They occur in areas without bare cover and have a relatively high percentage of
287 cropland (over 20%).

288 *3.1.2 Cropland systems*

289 Cropland systems cover 37.8% of the region, significantly higher than the estimated global
290 average of 8% (van Asselen & Verburg, 2012). This makes them the most represented land
291 system group in the Mediterranean region. They are defined by a high average of cropland cover

292 of over 45% but also contain significant portions of tree and bare cover. Cropland systems are
293 divided into three categories: extensive, intensive rainfed and irrigated, and are further
294 subdivided into annual and permanent crop systems, and mosaics of annual and permanent crops.
295 Extensive systems cover vast areas in North Africa, the Middle East and the Anatolian plateau in
296 Turkey. Intensive rainfed cropland systems mostly occur in the Northern Mediterranean (Spain,
297 Italy, France, parts of Turkey) with the notable exception of northern Tunisia. Irrigated systems
298 occur throughout the region, often along major rivers (Nile in Egypt, Euphrates and Tigris in
299 Turkey and Syria, Guadalquivir in Spain, Sebou and Sous in Morocco.

300 *3.1.3 Forest systems*

301 The global estimate for forest systems is 21% of the global surface (van Asselen & Verburg,
302 2012), whereas in the Mediterranean region we estimate these systems to cover 10.1%. Forest
303 systems are characterized by a high, over 40% average tree cover. Notable portions of areas with
304 higher tree density are however represented as agro-silvo-pastoral mosaic systems (e.g. closed
305 wooded rangelands). Forest systems together with such dense tree cover mosaic systems cover
306 25.2% of the Mediterranean region. More than half of all forest areas are thus used for
307 cultivation and grazing. Most of the forests are in the mountainous regions of the European
308 Mediterranean. In the MENA region, continuous forest systems are situated in the Atlas
309 mountains spanning from Morocco to Tunisia (Fig. 5). Extensive areas covered by
310 Mediterranean forest systems occur on Corsica, the most forested Mediterranean island (Fig. 4b).
311 Most of the forests are defined by medium intensity management (61.1%), followed by natural
312 and semi-natural forests (25.5%). A lower extent of forests is characterized by high intensity
313 management (10%) or as planted forests (3.4%), mostly occurring on the Iberian peninsula.

314 3.1.4 *Agro-silvo-pastoral mosaics*

315 Mosaic systems cover 23.3% of the Mediterranean - this is substantially higher compared to the
316 4–9% global estimates of mosaic cropland, grassland and forest systems (van Asselen &
317 Verburg, 2012). They are characterized by a medium to high average cropland cover (14 to
318 60%), and hold a considerable portion of areas covered by tree cover. The four
319 woodland/wooded rangeland classes, would be represented as forest cover in an approach
320 focusing on dominant land cover. In this study, they however represent landscapes, where forest
321 activities coincide with grazing and arable cultivation. The open woodland class represents areas
322 with moderate average tree cover (17.2%) and a lower livestock density (31.2 animals/km²).
323 Open wooded rangelands have a similar average tree cover (16.0%), however a higher average
324 livestock density (84.5 animals/km²). The cropland and wooded rangeland mosaic systems are
325 also defined by a high average cropland cover of 39.0%. All three open woodland systems occur
326 in the whole Mediterranean region, with the most notable examples of the dehesa/montado
327 system of the Iberian peninsula (Fig. 4c). Closed wooded rangeland are limited to areas in the
328 Atlas mountains, Albania and Greece, Sicily, Sardinia and central Spain. They have a high
329 average tree cover (38.5%) and a high average livestock density (98.5 animals/km²). In the
330 remaining two systems, crop cultivation and livestock grazing occurs on the same space. The
331 cropland and rangeland system mostly are mostly low-intensity cereal fields with livestock
332 grazing. Such systems are present on vast areas in North-West Africa, the Iberian peninsula, the
333 Anatolian plateau in Turkey and in the Middle East. The permanent crops and rangeland systems
334 are present in Syria, Tunisia and Morocco (Fig. 5).

335

336 3.1.5 Settlement systems

337 Settlement systems occupy 5.4% of the Mediterranean region, with 4.1% being attributed to peri-
338 urban areas, and 1.3% to urban areas. These systems have a high share of cropland cover (46 and
339 32% respectively), and high livestock density (78 and 51 animals/km² respectively). Most urban
340 systems are found along the Mediterranean coastline, with few notable exceptions situated on the
341 mainland (Amman, Ankara, Marrakesh, Madrid, etc.).

342 3.1.6 Wetlands

343 Wetland systems represent lakes and other wetlands that are not managed as irrigated cropland.
344 Wetland systems are characterized by a high average value of bare cover (38.3%). Extensive salt
345 lakes occur in the desert regions of North Africa, known as “chotts” or “sebkhas”. Often they are
346 seasonal wetlands that dry out in the summer (Khaznadar *et al.*, 2009), and are represented as
347 deserts in land cover products. Wetlands in the Mediterranean also have a high average livestock
348 density of 353 animals/km². Historically, wetlands in the MENA region have been a source of
349 water and fodder for livestock, with numbers of livestock grazing still increasing (Houérou,
350 1993; Médail & Quézel, 1999).

351 3.2 Location factors

352
353 The results of the binominal logistic regression are summarized in Table 3 and Supplement D.
354 Overall, we see high fits of the regression models, indicating that the selected location factors
355 can explain a large fraction of the spatial variation in occurrence of the different land systems.

356

357 *3.2.1 Bare and open grazing systems*

358 Bare and open grazing systems generally occur in remote areas with a lower population density -
359 with the exception of the intensive arid grazing system, that tends to occur close to markets. This
360 system tends to occur in areas with higher solar radiation and lower potential evapotranspiration
361 (PET). The two arid grazing systems occur in areas with low precipitation and their likelihood
362 increases with rising altitudes.

363 *3.2.2 Cropland systems*

364 Cropland systems occur in areas with lower altitudes and gentle slopes. Temperature has a
365 positive association with most cropland systems. Although these systems tend to be negatively
366 related to population density, irrigated systems occur in areas with higher density of rural
367 population. The location of these systems is positively related with market influence. This can
368 be explained by the investments in the agricultural sector and the potential to sell products,
369 which is possible in areas with a high market influence. Soil pH levels have a positive influence
370 on the occurrence of cropland systems, whereas the soil organic content is negatively related to
371 their occurrence.

372 *3.2.3 Forest systems*

373 Forest systems tend to be negatively related to population density. These systems are positively
374 related to soil sand content, and negatively to pH levels and soil depth. When using soil
375 characteristics based on current land cover, forest systems were positively related to organic
376 content and soil depth. Clearly, to some extent these environmental conditions are a result of the
377 influence of the forest ecosystem on the soil conditions itself. Forests are more frequently found
378 on slopes and in areas with higher precipitation (except planted forests). Mediterranean natural

379 and semi-natural forests are positively related to altitude and temperature. Planted forests are
380 positively related to well-drained soils. While Mediterranean planted forests can consist of native
381 species well adapted to aridity, young plantations of introduced species such as the Monterey
382 pine (*Pinus radiata*) have higher water demands and prefer well drained soils (Garmendia *et al.*,
383 2012).

384 *3.2.4 Agro-silvo-pastoral mosaics*

385 Although mosaic systems have very different characteristics amongst the sub-types, they do have
386 some similarities. They tend to be negatively related to population density, soil pH and soil
387 depth. The cropland/rangeland categories have similar characteristic as cropland systems in
388 terms of relation to slope, and have a positive association with potential evapotranspiration like
389 intensive cropland systems. The woodland/wooded rangeland categories are similar to forest
390 systems in terms of relations to soils characteristics, as well as to slope and precipitation. The
391 results show that agro-silvo-pastoral mosaics resemble either cropland or forests systems in
392 terms of location specific characteristics. This is logical, as they are either croplands, or
393 woodlands, where other activities occur on the same space.

394 *3.2.5 Settlement systems and wetlands*

395 Settlement systems are, almost by definition, positively related to population density,
396 infrastructure and market accessibility. They occur on lower altitudes with gentler slopes.
397 Wetlands occur on flat areas with lower altitudes, and have a negative association with
398 temperature, population density and market influence.

399

400 3.3 Performance and data uncertainty

401
402 Studies used in the validation covered the whole region (Fig. 7), although more were found in the
403 European part (122) as compared to the MENA region (68). Out of 190 documented studies, 134
404 had perfect agreement (71%), 42 partial agreement (22%), and 14 were misclassified (7%),
405 compared to our map. Studies with partial agreement had a correct identification of the land
406 systems group, however a different land systems subgroup. The accuracies of aggregated land
407 system categories shows the extent of inter-category misclassifications and complete
408 misclassifications (Table 4). The producer's accuracy presents the extent of how well the
409 documented land systems were represented on the land systems map. The user's accuracy also
410 takes into account the extent of land systems attributed to other systems. Interestingly, our user's
411 and producer's accuracies are in a similar range as is common for remote sensing interpretations
412 of land cover.

413 Using only data with global coverage to produce the land systems map shows the drawbacks of
414 using such data. It is difficult to differentiate between systems of different intensities and type of
415 crops if only using proxies for intensity (Fig. 8, Supplement E). The differences are smaller for
416 systems classified with data on bare areas, irrigation, livestock and tree cover.

417 When using global data, urban and peri-urban systems in the European part are overestimated
418 (Fig. 8). All agro-silvo-pastoral mosaic systems have a low agreement between the maps,
419 indicating that using data with global coverage significantly underestimates these areas. Mosaic
420 systems are mostly lost on the account of more intensive cropland systems. Extensive annual
421 cropland and all three annual-permanent mosaic systems cover significantly more areas, with
422 permanent crop systems experiencing substantial losses. The changes are not only in terms of

423 quantities of such systems, but mostly in their allocation, leading to a different spatial pattern
424 (Fig. 8). Vast areas in Europe lose the fine detailed structure of cropland and agro-silvo-pastoral
425 systems, and are represented by areas where both annual and permanent crops are cultivated
426 (Supplement E).

427 **4. Discussion**

428 **4.1 Classifying Mediterranean land systems**

429
430 Representing the spatial pattern and intensity of human-environment interactions remains one of
431 the most significant challenges in land systems science (Rounsevell *et al.*, 2012; Turner *et al.*,
432 2007). Several authors have previously combined data to improve information on land use and
433 management. Global scale land system characterizations include those of Ellis and Ramankutty
434 (2008) who mapped anthropogenic biomes using numerous socio-economic and bio-physical
435 indicators. Van Asselen and Verburg (2012) mapped global land systems, and investigated their
436 spatial determinants. Letourneau *et al.* (2012) classified land-use systems for use in the context
437 of the integrated assessment model IMAGE. Václavík *et al.* (2013) classified land system
438 archetypes based on similarities in a broad range of characteristics. Although recognizing similar
439 systems on a global scale is useful for global assessments and modeling, these approaches fail to
440 capture the diverse regional characteristics and do not always link to local systems and
441 nomenclatures (Václavík *et al.*, 2013). On the other end of the spectrum are farming system
442 classifications operating at the farm level, ignoring the larger landscape context, which is
443 important for many of the services provided by these systems (Dixon *et al.*, 2001; van de Steeg
444 *et al.*, 2010). Regional scale characterizations were made by Levers *et al.* (2015) and van der
445 Zanden *et al.* (2016), mapping land system archetypes and cultural landscapes of Europe

446 respectively. Levers et al. (2015) generalized Mediterranean mosaic archetypes to low intensity
447 cropland, grassland or mosaic systems, grouping them together with low intensity single function
448 systems. In the study of van der Zanden et al. (2016), several mosaic landscape types of different
449 intensities were identified, however disregarding woodland systems. Our approach moved
450 beyond existing classification systems by accounting for the specific land systems characteristic
451 for the Mediterranean region. We identified 6 agro-silvo-pastoral classes that are all, functionally
452 different, variations of mosaic land systems. Although the value of these mosaic systems for
453 society and biodiversity is known, this is the first time their spatial extent and pattern is mapped.

454 Thresholds used in our classification are often difficult to identify and are to some extent
455 arbitrary. For example, classifying different grazing systems is challenging, as transhumance is
456 still significant in the Mediterranean region – livestock may only be present in an area during a
457 particular time of the year. Sheep densities on barley fields might increase to 65 animals/ha for
458 one month each year, in order to supplement the animals' summer diet (Correal *et al.*, 2006). In
459 traditional continuous forage systems livestock densities are much lower, with up to 2 animals/ha
460 (Delgado *et al.*, 2004). We focused on such systems, and did not include the temporal variability
461 of livestock. Another example are forest systems, defined as land with over 10% tree cover by
462 the FAO (FAO, 2000). This definition includes significant areas of woodlands hosting mosaic
463 systems.

464 **4.2 Uncertainties in data**

465
466 Significant improvements have been made in providing global data on land cover and
467 management intensity. Nevertheless, there are still considerable inconsistencies between
468 different global data sets contributing to the data uncertainty (Tuanmu & Jetz, 2014). Combining

469 different data sets derived from remote sensing, modeling or censuses can result in aggregating
470 the inaccuracies of those data sets. As fully harmonized data on the different aspects are not
471 available, the possible bias from inconsistencies between the different data layers is unavoidable.
472 Sometimes, these inconsistencies reveal interesting information. We observed that the European
473 sealed soil map defined protected agricultural areas (greenhouses) in south of Spain as sealed
474 surfaces. This resulted in a misclassification of both the cropland and urban classes in this
475 particular area. Although protected agriculture could be defined as a sealed surface, the same
476 error does not occur in other regions with vast areas of protected agriculture (Greece, Italy). This
477 prevented us from identifying protected agriculture as a separate land system using the
478 combination of sealed or urban areas with cropland extent. Spatially explicit data on protected
479 agriculture in the region is basically non-existent and is limited to a few areas in Italy, Israel and
480 Spain (Aguilar *et al.*, 2015; Levin *et al.*, 2007; Picuno *et al.*, 2011).

481 Additional data related issues are the over- and underrepresentation of particular systems.
482 Despite the good coverage of high resolution remote sensing derived products (Hansen *et al.*,
483 2013), areas covered by forests are underrepresented in the MENA region. Our analysis has
484 shown a potential overestimation of intensive, and underestimation of mosaic land systems in the
485 data poor parts of the Mediterranean (Fig. 8, Supplement E).

486 In terms of agricultural and forest management intensity, there is inadequate global data, or it is
487 not available at sufficiently detailed spatial resolution (Hurt *et al.*, 2006; Ramankutty *et al.*,
488 2008). To identify the intensity of Mediterranean land systems, we had to use a set of different
489 proxies. Our combination of field size and yield used in the non-European part of the region did
490 not consider the numerous aspects of both the input and output intensities (Erb *et al.*, 2013).
491 Yields and management are varying with time and incorporating multi-temporal data could

492 improve the identification of management intensity (Levers *et al.*, 2015). Similar concerns hold
493 for forest management. Although we used temporal changes in forest cover as a proxy for forest
494 management for the non-European Mediterranean part, other data such as wood production and
495 socio-economic statistics could be helpful (Verkerk *et al.*, 2015).

496 This study presents a novel data assimilation approach to identify the extent and spatial patterns
497 of Mediterranean land systems. As land systems are composed of different components, their
498 characteristics will never be measured and observed by single sensors. Combining different
499 datasets will, therefore, always be needed to update the map in the future.

500 **4.3 Application of results**

501
502 The resulting land systems map has a wide potential of use. The identified extent of agro-silvo-
503 pastoral mosaics can be used for prioritization of landscapes for biodiversity and cultural
504 heritage conservation. The results can also be used in earth system modeling, as using land
505 systems in such models can provide a more accurate representation of the intensity of human-
506 environment interactions (van Asselen & Verburg, 2012). When modeling climate impacts, using
507 such a map can provide more information. For example, the albedo and greenhouse gas
508 emissions and sequestration will be different between the systems. The results can also be used
509 in land-change models or in integrated assessment models, to analyze consequences of future
510 socio-economic changes (Verburg *et al.*, 2011a). Using land systems we can capture changes in
511 management intensity, as socio-economic changes often do not affect land cover directly.

512 To improve our approach, better data is needed for the Middle Eastern and North African part of
513 the region. Vast areas of extensive cropland and agro-silvo-pastoral mosaic systems are present

514 there, significant for regional food security and biodiversity. These areas are also more
515 representative for other cropland and woodland areas in semi-arid regions.

516 **5. Conclusion**

517 Mediterranean landscapes have been shaped through centuries by human activities in often harsh
518 environmental conditions. This has resulted in diverse land systems with high cultural values and
519 of high importance for regional food production. Our typology provides a first map that
520 represents diverse land systems, including multifunctional landscapes and other aspects of land
521 management in the Mediterranean region that have been widely studied but not represented in
522 maps. This typology helps to improve the understanding of Mediterranean land systems and is a
523 basis for assessments of future changes in regional climate, land use and land cover change and
524 changes in management intensity. Compared to existing global and regional classifications our
525 typology significantly improved the thematic resolution and particularly was able to represent
526 agro-silvo-pastoral mosaic systems, which were mostly represented as single function low
527 intensity grassland or cropland in other studies. The comparison with case studies throughout the
528 region has shown that our map sufficiently well represents the variation in land systems across
529 the region and, thus, can be used to support prioritization of areas for biodiversity protection,
530 conservation of cultural landscapes, or food production.

References

- Agnoletti M (2014) Rural landscape, nature conservation and culture: Some notes on research trends and management approaches from a (southern) European perspective. *Landscape and Urban Planning*, **126**, 66–73. doi - 10.1016/j.landurbplan.2014.02.012
- Aguilar MA, Vallario A, Aguilar FJ, Lorca AG, Parente C (2015) Object-Based Greenhouse Horticultural Crop Identification from Multi-Temporal Satellite Imagery: A Case Study in Almeria, Spain. *Remote Sensing*, **7**, 7378–7401. doi - 10.3390/rs70607378
- Almagro M, Vente J de, Boix-Fayos C et al. (2013) Sustainable land management practices as providers of several ecosystem services under rainfed Mediterranean agroecosystems. *Mitigation and Adaptation Strategies for Global Change*, 1–15. doi - 10.1007/s11027-013-9535-2
- Almeida M, Guerra C, Pinto-Correia T (2013) Unfolding relations between land cover and farm management: high nature value assessment in complex silvo-pastoral systems. *Geografisk Tidsskrift-Danish Journal of Geography*, **113**, 97–108. doi - 10.1080/00167223.2013.848611
- van Asselen S, Verburg PH (2012) A Land System representation for global assessments and land-use modeling. *Global Change Biology*, **18**, 3125–3148. doi - 10.1111/j.1365-2486.2012.02759.x
- Blanchard SD, Pontius Jr. RG, Urban KM (2015) Implications of Using 2 m versus 30 m Spatial Resolution Data for Suburban Residential Land Change Modeling. *Journal of Environmental Informatics*. doi - 10.3808/jei.201400284
- Britz W, Witzke P (2014) *Common Agricultural Policy regionalised Impact Modelling System (CAPRI). Modelling System Documentation*. Bonn.
- Brus DJ, Hengeveld GM, Walvoort DJJ, Goedhart PW, Heidema AH, Nabuurs GJ, Gunia K (2012) Statistical mapping of tree species over Europe. *European Journal of Forest Research*, **131**, 145–157. doi - 10.1007/s10342-011-0513-5
- Bunce RGH, Pérez-Soba M, Smith M (2008) Assessment of the Extent of Agroforestry Systems in Europe and Their Role Within Transhumance Systems. In: *Agroforestry in Europe: Current Status and Future Prospects* (eds Rodríguez AR, McAdam J, Mosquera-Losada MR). Springer Science & Business Media.
- CIESIN (2015) Gridded Population of the World, Version 4. Center for International Earth Science Information Network - CIESIN - Columbia University. NASA Socioeconomic Data and Applications Center (SEDAC).
- CIESIN, IFPRI, CIAT (2011) Global Rural-Urban Mapping Project, Version 1: Population Density Grid. Center for International Earth Science Information Network - CIESIN - Columbia University, International Food Policy Research Institute - IFPRI, The World Bank, and Centro Internacional de Agricultura Tropical - CIAT.
- Correal E, Robledo A, Rios S, Rivera D (2006) Mediterranean dryland mixed sheep-cereal systems. *Grassland Science in Europe*, **11**, 14–26.
- Cuttelod A, Garcia N, Malak DA, Temple HJ, Katariya V (2009) The Mediterranean: a biodiversity hotspot under threat. In: *Wildlife in a Changing World: An Analysis of the 2008 IUCN Red List of Threatened Species* (eds Vie J-C, Hilton-Taylor C, Stuart SN). International Union for Conservation of Nature, Gland, Switzerland.
- Daccache A, Ciurana JS, Diaz JAR, Knox JW (2014) Water and energy footprint of irrigated agriculture in the Mediterranean region. *Environmental Research Letters*, **9**, 124014. doi - 10.1088/1748-9326/9/12/124014
- van Delden H, van Vliet J, Rutledge DT, Kirkby MJ (2011) Comparison of scale and scaling issues in integrated land-use models for policy support. *Agriculture, Ecosystems & Environment*, **142**, 18–28. doi - 10.1016/j.agee.2011.03.005
- Delgado I, Andueza D, Muñoz F, Lahoz F (2004) Forage system to replace marginal rainfed cereal areas by sheep production. An experimental study. *Options Méditerranéennes. Serie A*, **60**, 263–266.

- Dixon J, Gulliver A, Gibbon D, Hall M (2001) *Farming Systems and Poverty: Improving Farmers' Livelihoods in a Changing World*. FAO and World Bank, Rome and Washington D.C.
- EEA (2015a) CORINE Land Cover 2006 — Copernicus Land Monitoring Services.
- EEA (2015b) Imperviousness Soil Sealing High Resolution Data. Copernicus Land Monitoring Services. European Environment Agency.
- Ellis EC, Ramankutty N (2008) Putting people in the map: anthropogenic biomes of the world. *Frontiers in Ecology and the Environment*, **6**, 439–447. doi - 10.1890/070062
- Erb K-H, Haberl H, Jepsen MR et al. (2013) A conceptual framework for analysing and measuring land-use intensity. *Current Opinion in Environmental Sustainability*, **5**, 464–470. doi - 10.1016/j.cosust.2013.07.010
- Evans JP (2008) 21st century climate change in the Middle East. *Climatic Change*, **92**, 417–432. doi - 10.1007/s10584-008-9438-5
- Fader M, Shi S, von Bloh W, Bondeau A, Cramer W (2016) Mediterranean irrigation under climate change: more efficient irrigation needed to compensate increases in irrigation water requirements. *Hydrology and Earth System Sciences*, **20**, 953–973. doi - 10.5194/hess-20-953-2016
- FAO (2000) *Global Forest Resource Assessment 2000 – Main Report, FAO Forestry Paper 140*. United Nations Food and Agricultural Organization, Rome.
- FAO (2011) *Land Degradation Assessment in Drylands. Mapping Land Use Systems at Global and Regional Scales for Land Degradation Assessment Analysis. Version 1.1*. FAO, Rome.
- Fritz S, See L, McCallum I et al. (2015) Mapping global cropland and field size. *Global Change Biology*, **21**, 1980–1992. doi - 10.1111/gcb.12838
- Fuchs R, Herold M, Verburg PH, Clevers JGPW (2013) A high-resolution and harmonized model approach for reconstructing and analysing historic land changes in Europe. *Biogeosciences*, **10**, 1543–1559. doi - 10.5194/bg-10-1543-2013
- García-Llorente M, Martín-López B, Iniesta-Arandia I, López-Santiago CA, Aguilera PA, Montes C (2012) The role of multi-functionality in social preferences toward semi-arid rural landscapes: An ecosystem service approach. *Environmental Science & Policy*, **19–20**, 136–146. doi - 10.1016/j.envsci.2012.01.006
- Garmendia E, Mariel P, Tamayo I, Aizpuru I, Zabaleta A (2012) Assessing the effect of alternative land uses in the provision of water resources: Evidence and policy implications from southern Europe. *Land Use Policy*, **29**, 761–770. doi - 10.1016/j.landusepol.2011.12.001
- Guerra CA, Metzger MJ, Maes J, Pinto-Correia T (2015) Policy impacts on regulating ecosystem services: looking at the implications of 60 years of landscape change on soil erosion prevention in a Mediterranean silvo-pastoral system. *Landscape Ecology*, 1–20. doi - 10.1007/s10980-015-0241-1
- Hansen MC, Potapov PV, Moore R et al. (2013) High-Resolution Global Maps of 21st-Century Forest Cover Change. *Science*, **342**, 850–853. doi - 10.1126/science.1244693
- Hengeveld GM, Nabuurs G-J, Didion M, van den Wyngaert I, Clerkx APPM, Schelhaas M-J (2012) A Forest Management Map of European Forests. *Ecology and Society*, **17**. doi - 10.5751/ES-05149-170453
- Hengl T, de Jesus JM, MacMillan RA et al. (2014) SoilGrids1km — Global Soil Information Based on Automated Mapping. *PLoS ONE*, **9**, e105992. doi - 10.1371/journal.pone.0105992
- Hijmans RJ, Cameron SE, Parra JL, Jones PG, Jarvis A (2005) Very high resolution interpolated climate surfaces for global land areas. *International Journal of Climatology*, **25**, 1965–1978. doi - 10.1002/joc.1276
- Houérou HNL (1993) Salt-tolerant plants for the arid regions of the Mediterranean isoclimatic zone. In: *Towards the rational use of high salinity tolerant plants* (eds Lieth H, Masoom AAA), pp. 403–422. Springer Netherlands. doi - 10.1007/978-94-011-1858-3_42
- Huld T, Müller R, Gambardella A (2012) A new solar radiation database for estimating PV performance in Europe and Africa. *Solar Energy*, **86**, 1803–1815. doi - 10.1016/j.solener.2012.03.006

- Hurttt GC, Frolking S, Fearon MG et al. (2006) The underpinnings of land-use history: three centuries of global gridded land-use transitions, wood-harvest activity, and resulting secondary lands. *Global Change Biology*, **12**, 1208–1229. doi - 10.1111/j.1365-2486.2006.01150.x
- IUCN (2015) World Database on Protected Areas.
- Joffre R, Rambal S, Ratte JP (1999) The dehesa system of southern Spain and Portugal as a natural ecosystem mimic. *Agroforestry Systems*, **45**, 57–79. doi - 10.1023/A:1006259402496
- Jun C, Ban Y, Li S (2014) China: Open access to Earth land-cover map. *Nature*, **514**, 434–434. doi - 10.1038/514434c
- Khaznadar M, Vogiatzakis IN, Griffiths GH (2009) Land degradation and vegetation distribution in Chott El Beida wetland, Algeria. *Journal of Arid Environments*, **73**, 369–377. doi - 10.1016/j.jaridenv.2008.09.026
- Kuemmerle T, Erb K, Meyfroidt P et al. (2013) Challenges and opportunities in mapping land use intensity globally. *Current Opinion in Environmental Sustainability*, **5**, 484–493. doi - 10.1016/j.cosust.2013.06.002
- Lambin EF, Turner BL, Geist HJ et al. (2001) The causes of land-use and land-cover change: moving beyond the myths. *Global Environmental Change*, **11**, 261–269. doi - 10.1016/S0959-3780(01)00007-3
- Latham J, Cumani R, Rosati I, Bloise M (2014) *Global Land Cover SHARE Database Beta-Release Version 1.0*. FAO, Rome.
- Letourneau A, Verburg PH, Stehfest E (2012) A land-use systems approach to represent land-use dynamics at continental and global scales. *Environmental Modelling & Software*, **33**, 61–79. doi - 10.1016/j.envsoft.2012.01.007
- Levers C, Müller D, Erb K et al. (2015) Archetypical patterns and trajectories of land systems in Europe. *Regional Environmental Change*, 1–18. doi - 10.1007/s10113-015-0907-x
- Levin N, Lugassi R, Ramon U, Braun O, Ben-Dor E (2007) Remote sensing as a tool for monitoring plasticulture in agricultural landscapes. *International Journal of Remote Sensing*, **28**, 183–202. doi - 10.1080/01431160600658156
- Médail F, Quézel P (1999) Biodiversity Hotspots in the Mediterranean Basin: Setting Global Conservation Priorities. *Conservation Biology*, **13**, 1510–1513. doi - 10.1046/j.1523-1739.1999.98467.x
- Meeus JHA (1995) Pan-European landscapes. *Landscape and Urban Planning*, **31**, 57–79. doi - 10.1016/0169-2046(94)01036-8
- Neumann K, Sietz D, Hilderink H, Janssen P, Kok M, van Dijk H (2015) Environmental drivers of human migration in drylands – A spatial picture. *Applied Geography*, **56**, 116–126. doi - 10.1016/j.apgeog.2014.11.021
- NGIA (2015) VMap0 data. National Geospatial Intelligence Agency. <http://gis-lab.info/qa/vmap0-eng.html>.
- Nieto-Romero M, Oteros-Rozas E, González JA, Martín-López B (2014) Exploring the knowledge landscape of ecosystem services assessments in Mediterranean agroecosystems: Insights for future research. *Environmental Science & Policy*, **37**, 121–133. doi - 10.1016/j.envsci.2013.09.003
- Olson DM, Dinerstein E, Wikramanayake ED et al. (2001) Terrestrial Ecoregions of the World: A New Map of Life on Earth A new global map of terrestrial ecoregions provides an innovative tool for conserving biodiversity. *BioScience*, **51**, 933–938.
- Panagos P, Hiederer R, Van Liedekerke M, Bampa F (2013) Estimating soil organic carbon in Europe based on data collected through an European network. *Ecological Indicators*, **24**, 439–450. doi - 10.1016/j.ecolind.2012.07.020
- Pfeiffer T, Koutantou A (2015) Greece debt crisis: Olive oil supplies threatened as banking system grinds to a halt. *The Independent*.
- Pickett STA, Cadenasso ML (1995) Landscape Ecology: Spatial Heterogeneity in Ecological Systems. *Science*, **269**, 331–334. doi - 10.1126/science.269.5222.331

- Picuno P, Tortora A, Capobianco RL (2011) Analysis of plasticulture landscapes in Southern Italy through remote sensing and solid modelling techniques. *Landscape and Urban Planning*, **100**, 45–56. doi - 10.1016/j.landurbplan.2010.11.008
- Plieninger T, Schaar M (2008) Modification of Land Cover in a Traditional Agroforestry System in Spain: Processes of Tree Expansion and Regression. *Ecology and Society*, **13**.
- Pontius RG, Santacruz A (2014) Quantity, exchange, and shift components of difference in a square contingency table. *International Journal of Remote Sensing*, **35**, 7543–7554. doi - 10.1080/2150704X.2014.969814
- Ramankutty N, Evan AT, Monfreda C, Foley JA (2008) Farming the planet: 1. Geographic distribution of global agricultural lands in the year 2000: Global Agricultural Lands in 2000. *Global Biogeochemical Cycles*, **22**. doi - 10.1029/2007GB002952
- Robinson TP, Wint GRW, Conchedda G et al. (2014) Mapping the Global Distribution of Livestock. *PLoS ONE*, **9**, e96084. doi - 10.1371/journal.pone.0096084
- Rodríguez C, Wiegand K (2009) Evaluating the trade-off between machinery efficiency and loss of biodiversity-friendly habitats in arable landscapes: The role of field size. *Agriculture, Ecosystems & Environment*, **129**, 361–366. doi - 10.1016/j.agee.2008.10.010
- Rounsevell MDA, Pedrolí B, Erb K-H et al. (2012) Challenges for land system science. *Land Use Policy*, **29**, 899–910. doi - 10.1016/j.landusepol.2012.01.007
- Schröter D, Cramer W, Leemans R et al. (2005) Ecosystem Service Supply and Vulnerability to Global Change in Europe. *Science*, **310**, 1333–1337. doi - 10.1126/science.1115233
- Siebert S, Döll P, Hoogeveen J, Faures J-M, Frenken K, Feick S (2005) Development and validation of the global map of irrigation areas. *Hydrology and Earth System Sciences*, **9**, 535–547. doi - 10.5194/hess-9-535-2005
- Siebert S, Henrich V, Frenken K, Burke J (2013) Update of the digital global map of irrigation areas to version 5. Rheinische Friedrich-Wilhelms-Universität, Bonn, Germany and FAO, Rome, Italy, 170pp.
- Sowers J, Vengosh A, Weinthal E (2010) Climate change, water resources, and the politics of adaptation in the Middle East and North Africa. *Climatic Change*, **104**, 599–627. doi - 10.1007/s10584-010-9835-4
- van de Steeg JA, Verburg PH, Baltenweck I, Staal SJ (2010) Characterization of the spatial distribution of farming systems in the Kenyan Highlands. *Applied Geography*, **30**, 239–253. doi - 10.1016/j.apgeog.2009.05.005
- Stoorvogel JJ, Bakkenes M, Temme AJAM, Batjes NH, ten Brink B (2016) S-World: a Global Soil Map for Environmental Modelling: S-World: a Global Soil Map for Environmental Modelling. *Land Degradation & Development*, **28**, 22–33. doi - 10.1002/ldr.2656
- Temme AJAM, Verburg PH (2011) Mapping and modelling of changes in agricultural intensity in Europe. *Agriculture, Ecosystems & Environment*, **140**, 46–56. doi - 10.1016/j.agee.2010.11.010
- Tuanmu M-N, Jetz W (2014) A global 1-km consensus land-cover product for biodiversity and ecosystem modelling. *Global Ecology and Biogeography*, **23**, 1031–1045. doi - 10.1111/geb.12182
- Turner BL, Lambin EF, Reenberg A (2007) The emergence of land change science for global environmental change and sustainability. *Proceedings of the National Academy of Sciences*, **104**, 20666–20671. doi - 10.1073/pnas.0704119104
- Václavík T, Lautenbach S, Kuemmerle T, Seppelt R (2013) Mapping global land system archetypes. *Global Environmental Change*, **23**, 1637–1647. doi - 10.1016/j.gloenvcha.2013.09.004
- Verburg PH, Soepboer W, Veldkamp A, Limpiada R, Espaldon V, Mastura SSA (2002) Modeling the Spatial Dynamics of Regional Land Use: The CLUE-S Model. *Environmental Management*, **30**, 391–405. doi - 10.1007/s00267-002-2630-x
- Verburg PH, Neumann K, Nol L (2011a) Challenges in using land use and land cover data for global change studies. *Global Change Biology*, **17**, 974–989. doi - 10.1111/j.1365-2486.2010.02307.x

- Verburg PH, Ellis EC, Letourneau A (2011b) A global assessment of market accessibility and market influence for global environmental change studies. *Environmental Research Letters*, **6**, 34019. doi - 10.1088/1748-9326/6/3/034019
- Verkerk PJ, Levers C, Kuemmerle T, Lindner M, Valbuena R, Verburg PH, Zudin S (2015) Mapping wood production in European forests. *Forest Ecology and Management*, **357**, 228–238. doi - 10.1016/j.foreco.2015.08.007
- Wright B, Cafiero C (2011) Grain reserves and food security in the Middle East and North Africa. *Food Security*, **3**, 61–76. doi - 10.1007/s12571-010-0094-z
- WWF (2004) Global Lakes and Wetlands Database. *World Wildlife Fund*.
- You L, Wood-Sichra U, Fritz S, Guo Z, See L, Koo L (2014) Spatial Production Allocation Model (SPAM) 2005 v2.0.
- Zamora J, Verdú JR, Galante E (2007) Species richness in Mediterranean agroecosystems: Spatial and temporal analysis for biodiversity conservation. *Biological Conservation*, **134**, 113–121. doi - 10.1016/j.biocon.2006.08.011
- van der Zanden EH, Levers C, Verburg PH, Kuemmerle T (2016) Representing composition, spatial structure and management intensity of European agricultural landscapes: A new typology. *Landscape and Urban Planning*, **150**, 36–49. doi - 10.1016/j.landurbplan.2016.02.005
- Zomer RJ, Trabucco A, Bossio DA, Verchot LV (2008) Climate change mitigation: A spatial analysis of global land suitability for clean development mechanism afforestation and reforestation. *Agriculture, Ecosystems & Environment*, **126**, 67–80. doi - 10.1016/j.agee.2008.01.014
- Zomer RJ, Trabucco A, Coe R, Place F (2009) *Trees on farm: analysis of global extent and geographical patterns of agroforestry*. World Agroforestry Centre, Nairobi, Kenya.

List of tables

Table 1. Data used in the hierarchical classification

Table 2: Location factors used in the regression analyses

Table 3: Regression coefficients for most significant Mediterranean land systems (full regression table in Supplement D; all coefficients significant at values below the 0.05 significance level)

Table 4: Shares of documented land system locations with perfect and partial agreement, and misclassification in %, together with the producer's and user's accuracy of the classification for aggregated land system categories

Table 1. Data used in the hierarchical classification

Group	Description	Original resolution	Spatial, temporal coverage	Unit	Source
Forest	Tree cover	30 m	Whole area, 2000 - 2014	%	Hansen et al. (2013)
	Tree cover gain and loss	30 m	Whole area, 2000 - 2014	presence	Hansen et al. (2013)
	Tree cover loss and gain ratio	30 m	Whole area, 2000 - 2014	%	Derived from Hansen et al. (2013)
	European forest management types	1 km	Europe, 2010	class	Hengeveld et al. (2012)
	Plantation tree species occurrence (<i>Eucalyptus</i> spp., <i>Populus</i> spp., <i>Pinus</i> spp., <i>Robinia</i> spp.)	1 km	Europe, 2000-2010	class	Brus et al. (2012)
Bare and artificial	Bare areas	30 m	Whole area, 2010	%	Latham et al. (2014)
	Built up areas	30 m	Whole area, 2010	%	Jun et al. (2014)
	Imperviousness	25 m	Europe, 2010	%	EEA (2015b)
Livestock and cropland	Cropland extent	1 km	Whole area, 2014	%	Fritz et al. (2015)
	Livestock density (bovines, goats and sheep)	1 km	Whole area, 2014	nr./ km ²	Robinson et al. (2014)
	Area equipped for irrigation	1 km	Whole area, 2006	ha	Siebert et al. (2013)
	European crop type map	vector	EU27, 2006	%	Britz and Witzke (2014)
	CORINE permanent crop land cover (vineyards, orchards, olive groves)	100 m	Non EU Europe and Turkey, 2006	class	EEA (2015a)
	SPAM permanent crop extent (oil, fruit, tropical fruit)	10 km	MENA, 2014	%	You et al. (2014)
	Fertilizer intensity	1 km	EU 27, 2000	class	Temme and Verburg (2011)
	Field size map	1 km	Whole area, 2015	class	Fritz et al. (2015)
	Areas with highest annual crop yield 10 th quantile of yields as intensification qualifies	MENA, Turkey	10 km, 2010	t/ha	You et al. (2014)
Areas with highest permanent crop yields (olives, temperate and tropic fruits) – 10 th quantile of yields as intensification qualifies	Non EU Europe, MENA, Turkey	10 km, 2010	t/ha	You et al. (2014)	
Other	Wetlands and lakes	250 m	Whole area, 2004	class	WWF (2004)
	Terrestrial ecoregions	vector	Whole area, 2001	class	Olson et al. (2001)
	Protected areas	vector	Whole area, 2001	class	IUCN (2015)

Table 2: Location factors used in the regression analyses

Location Factor	Unit/description	Resolution	Date	Source
Socio-economic				
Population density	People/km ²	1 km	2010	CIESIN (2015)
Rural population	Rural population/km ²	1 km	2000	CIESIN et al. (2011)
Market accessibility	Index (0-1)	1 km	2000-2010	Verburg et al. (2011b)
Market influence	USD/person (ppp)	1 km	2000-2010	Verburg et al. (2011b)
Accessibility	Distance to roads (m)	vector	1999	NGIA (2015)
Soil				
Drainage	Drainage class	1 km	2010	Hengl et al. (2014)
Sand content	Sand mass in %	1 km	2010	Stoorvogel (2016)
Clay content	Clay mass in %	1 km	2013	Stoorvogel (2016)
Cation Exchange Capacity (CEC)	cmol/kg	1 km	2010	Hengl et al. (2014)
pH	log(h+)	1 km	2010	Hengl et al. (2014)
Organic carbon content	g/kg in the top 50 cm	1 km	2013	Stoorvogel (2016)
Soil depth	cm	1 km	2013	Stoorvogel (2016)
Terrain				
Altitude	m above sea level	1 km	2005	Hijmans et al. (2005)
Slope	Slope degrees	1 km	2005	derived from Hijmans et al. (2005)
Climate				
Precipitation	annual precipitation (sum of monthly means) in mm	1 km	2005	Hijmans et al. (2005)
Temperature	Temperature (mean of monthly means) Celsius degree	1 km	2005	Hijmans et al. (2005)
Solar radiation	Horizontal surface irradiation (kWh/m ²), 1998-2011 mean	1.5 arc minute	2012	Huld et al. (2012)
Other				
Potential Evapotranspiration (PET)	annual PET in mm	1 km	2007	Zomer et al. (2008)
Potential vegetation	Pot. vegetation classes	10 km	2010	Ellis & Ramankutty (2008)

Table 3: Regression coefficients for most significant Mediterranean land systems (full regression table in Supplement D; all coefficients significant at values below the 0.05 significance level)

	Intensive arid grazing	Wetlands	Open wooded rangeland	Closed wooded rangeland	Extensive ann. cropland	Irrigated perm. crops	(semi) natural forest	Urban
Constant	-7.34	-0.38	-17.36	7.51	-1.20	-5.51	10.02	-0.20
Socio-economic								
Population density	-1.46E-3	-7.69E-4	-1.20E-3	-1.57E-3	-8.9E-4		-2.37E-3	3.86E-3
Rural population	-1.75E-3		-1.79E-3				-2.93E-3	-6.47E-3
Market accessibility	2.32		-3.11E-1	-4.40E-1	1.64	-2.30		1.53
Market influence	-2.50E-2	-1.57E-2			-1.735E-2	2.39E-2		
Road distance	-1.86E-5	2.00E-5			-3.08E-5	-8.64E-5		-2.90E-4
Soil characteristics								
Sand					1.91E-2		3.18E-2	
Clay		5.33E-2	-2.52E-2	-3.29E-2	3.32E-2			
CEC		5.34E-2	-1.90E-2		1.66E-2			
pH	1.63E-1		-3.39E-1		2.19E-1	2.80E-1	-3.00E-1	
Organic content	-5.36E-3	-8.42E-2			-8.57E-3	-7.35E-3		
Soil depth			-9.12E-3	-8.35E-3			-1.52E-2	
Drainage*		-1.50E-1 (b)						
Terrain								
Altitude	1.39E-3	-1.78E-3	-3.20E-4		3.90E-4	-6.72E-4	1.70E-3	-1.35E-3
Slope	7.64E-2	-4.88E-1	1.25E-1	1.25E-1		-1.93E-1	1.05E-1	-1.75E-1
Climate								
Precipitation	-2.91E-3		1.14E-3	2.85E-3	-2.56E-3			
Temperature		-1.38E-1		1.87E-1		4.59E-1	4.13E-1	
Solar radiation	4.79E-3	3.72E-3		-4.58E-3		-1.79E-3	-5.43E-3	
Other								
PET	-1.78E-3				-1.97E-3		-5.26E-3	
Potential natural vegetation**			-2.33 (7)		2.04 (2)		-4.47 (4)	
			-1.85 (9)		3.11 (4)		-4.09 (7)	
			3.98 (10)		1.99 (6)	-1.98 (7)	-2.02 (10)	
					2.59 (7)		4.68 (9)	
ROC	0.86	0.92	0.82	0.84	0.75	0.87	0.90	0.94

*Drainage classes: scale from a to g; a = poorly drained, g = excessively drained

**Potential natural vegetation: 2 = tropical deciduous woodland, 3 = temperate evergreen woodland, 6 = mixed woodland, 7 = savanna, 8 = grassland and steppe, 9 = dense shrubland, 10 = open shrubland

Table 4: Shares of documented land system locations with perfect and partial agreement, and misclassification in %, together with the producer's and user's accuracy of the classification for aggregated land system categories

Land system category	Perfect	Partial	Misclassification	Producer's accuracy	User's accuracy
Rangeland and grazing	56.5	26.1	17.4	66.7	88.9
Cropland	74.0	16.9	9.1	93.6	89.0
Forest	75.0	21.4	3.6	85.7	88.9
Agro-silvo-pastoral mosaics	69.2	28.9	1.9	84.6	87.8
(peri)Urban	100.0	0.0	0.0	100.0	61.5

List of figures

Fig. 1 Study area with the distribution of available high resolution (min. 1 km) spatial data layers (max. 15 data layers)

Fig. 2 Classification scheme. Detailed classification rules are presented in Supplement A.

Fig. 3 Distribution of Mediterranean land systems with locations of focus regions displayed in Fig. 4.

Fig. 4 Mediterranean land systems in more detail focusing on (a) Greece, (b) Sardinia and Corsica, (c) the Iberian peninsula, (d) Tunisia, (e) Morocco and (f) the Middle East

Fig. 5 Major Mediterranean land systems groups with percentage of total coverage of the Mediterranean ecoregion

Fig. 6 Average values of bare, cropland and tree cover, and livestock density per Mediterranean land system (Supplement C for detailed values and standard deviation)

Fig. 7 Locations and accuracy of the documented expert based validation

Fig. 8 Disagreement between the two land system map based on different data in terms of quantity and allocation

List of appendices

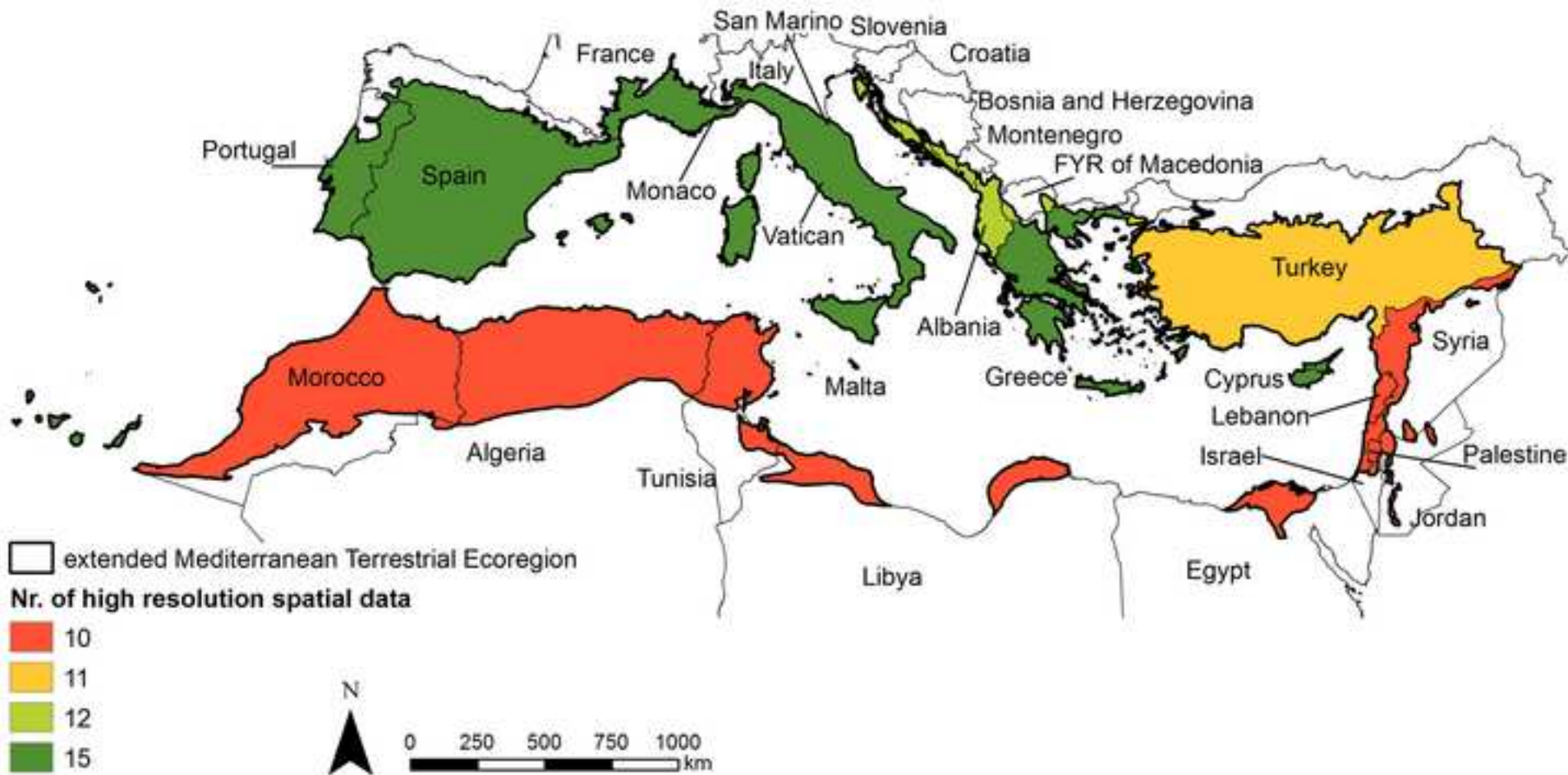
Supplement A: Classification details per land system

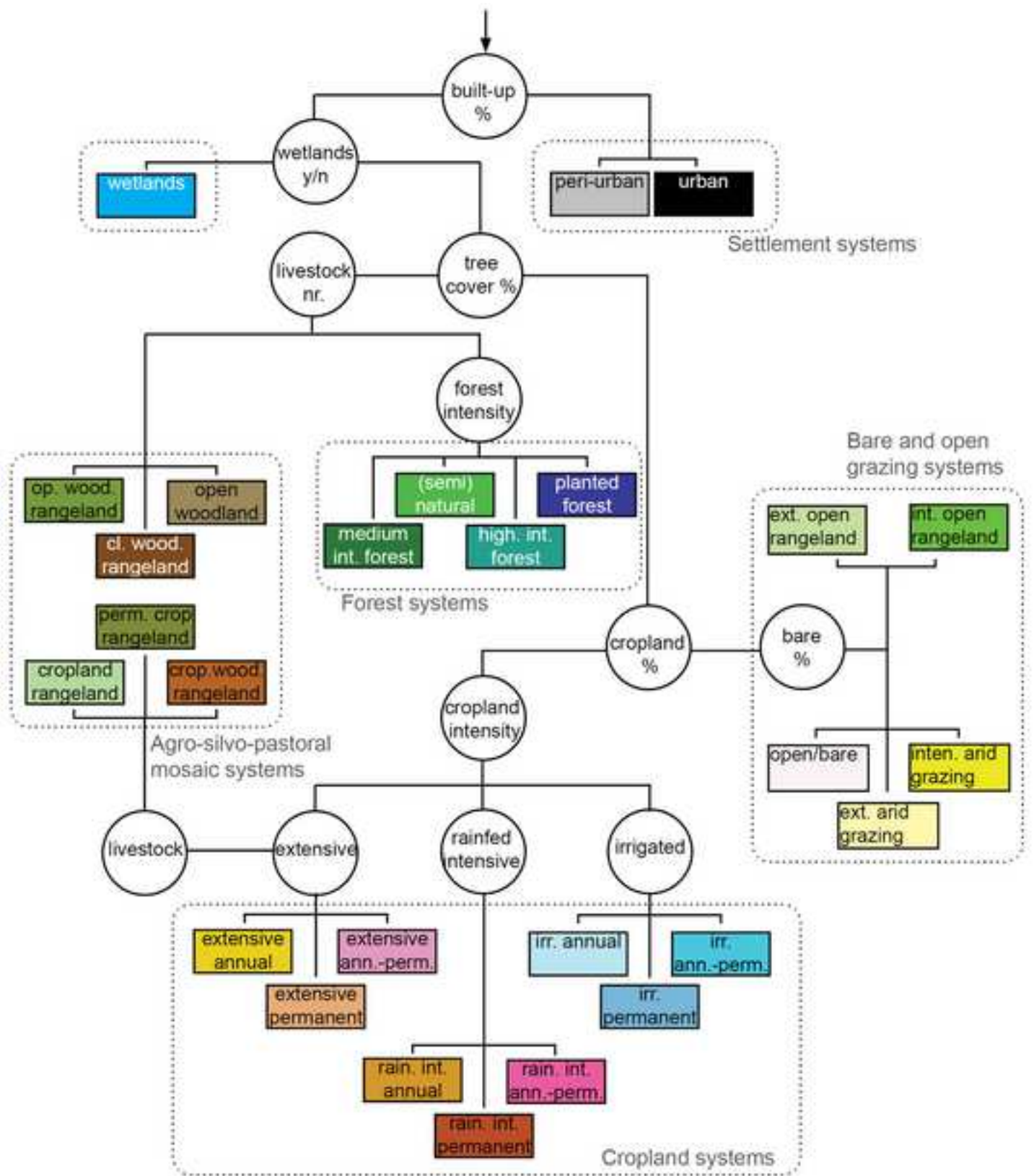
Supplement B: Studies used in the documented land systems validation per country

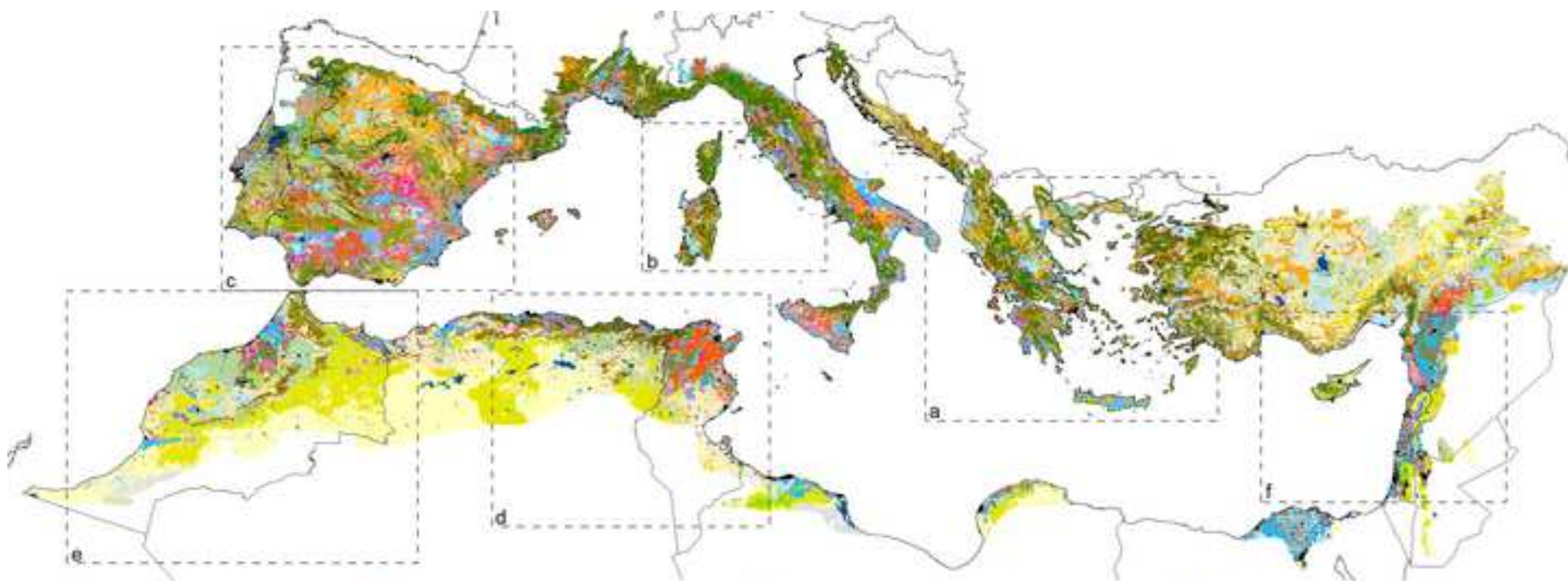
Supplement C: Average values for cropland, tree and bare cover, and livestock density per land system, standard deviations in brackets

Supplement D: Logistic regression results. All coefficients significant at values below the 0.05 significance level

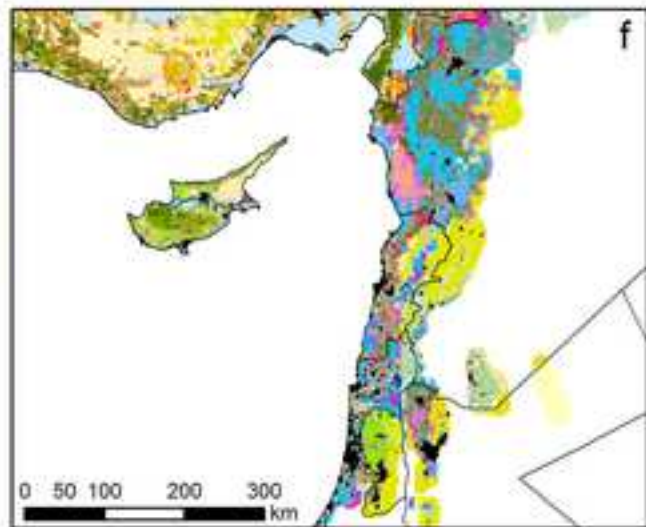
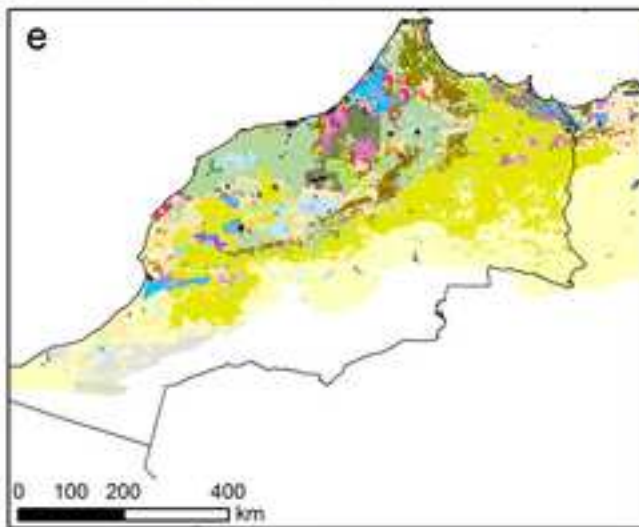
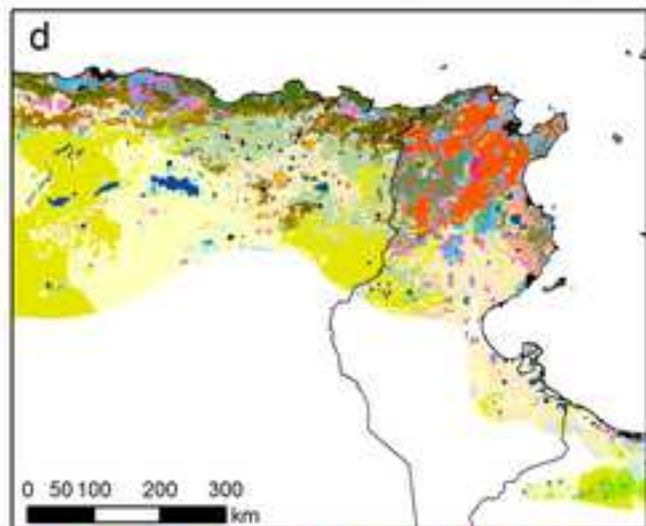
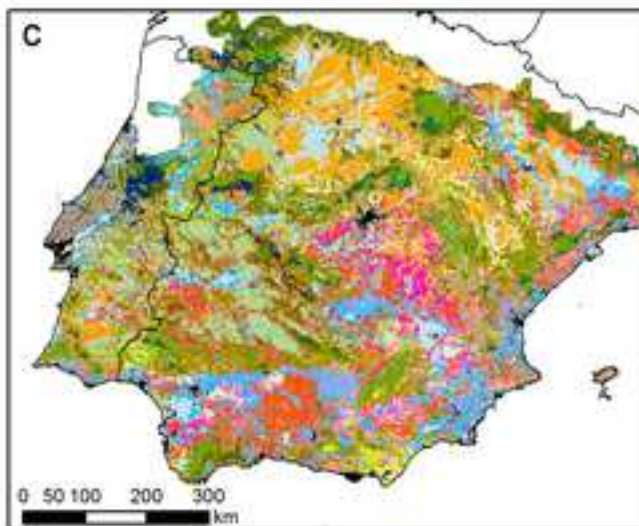
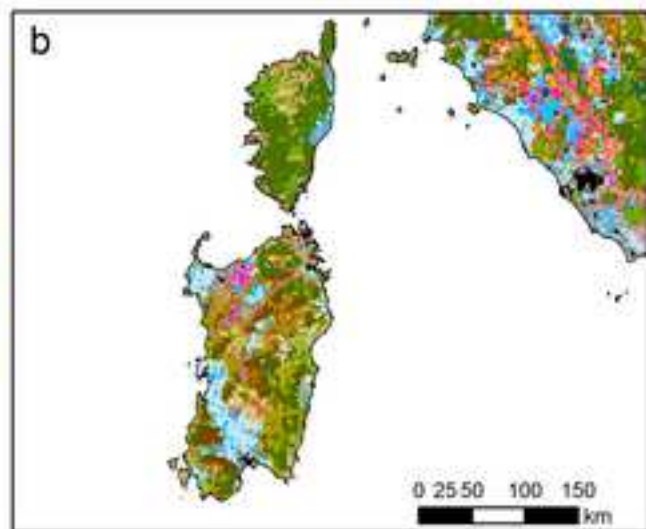
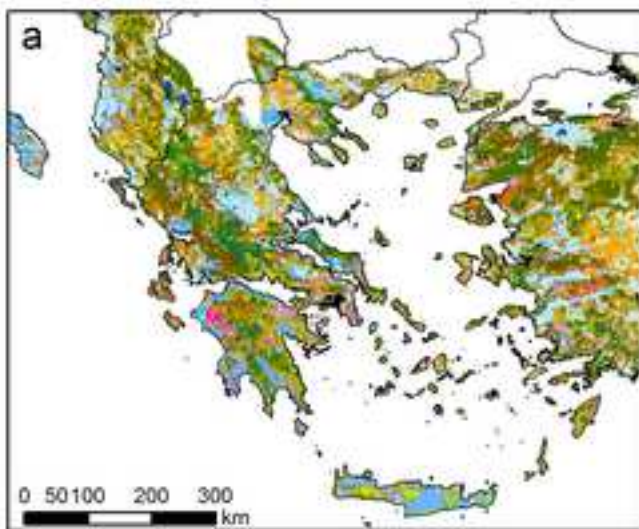
Supplement E: Mediterranean land systems map generated using globally available data







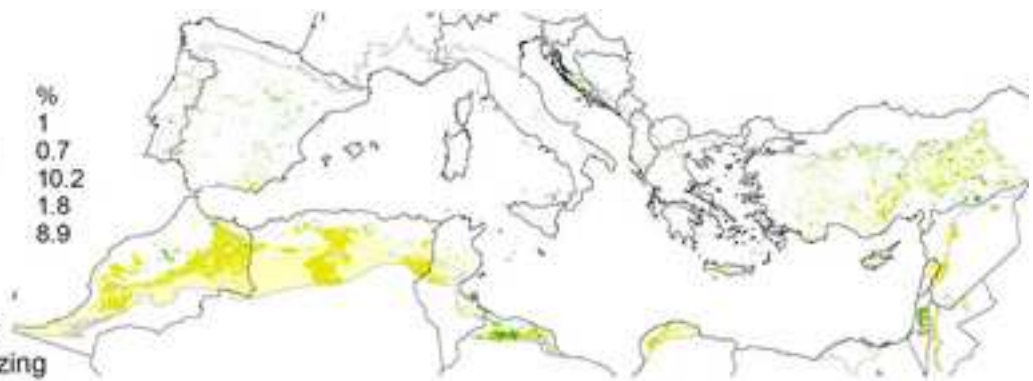
- | | | | |
|--------------------------|-------------------------------|-----------------------------------|---------------------------------|
| bare / desert | cropland and rangeland | extensive annual crops | medium intensity forest |
| open rangeland | open woodland | extensive permanent crops | natural and semi-natural forest |
| extensive arid grazing | open wooded rangeland | extensive ann-perm mosaic | high-intensity forest |
| intensive open rangeland | cropland and wooded rangeland | rainfed intensive annual crops | planted forests |
| intensive arid grazing | permanent crops and rangeland | rainfed intensive permanent crops | peri-urban |
| wetlands | closed wooded rangeland | rainfed intensive ann-perm mosaic | urban |
| | | irrigated annual crops | |
| | | irrigated permanent crops | |
| | | irrigated ann-perm mosaic | |



■ bare, desert	1
■ ext. open rangeland	0.7
■ ext. and grazing	10.2
■ int. open rangeland	1.8
■ int. arid grazing	8.9



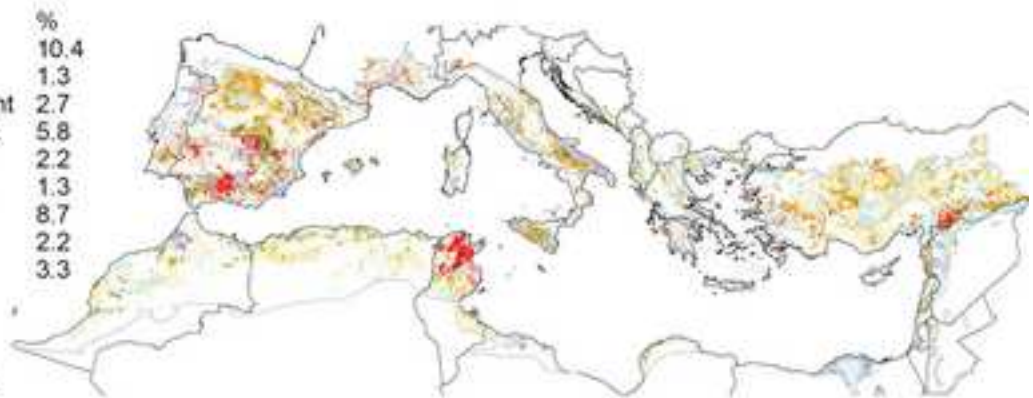
bare and open grazing systems



■ extensive annual	10.4
■ ext. permanent	1.3
■ ext. annual-permanent	2.7
■ rainfed intensive ann.	5.8
■ rain. int. permanent	2.2
■ rain. int. ann.-perm.	1.3
■ irrigated annual	8.7
■ irr. permanent	2.2
■ irr. ann.-perm.	3.3



cropland systems



■ medium intensity forest	6.2
■ semi(natural)	2.6
■ high intensity	1
■ planted forests	0.3



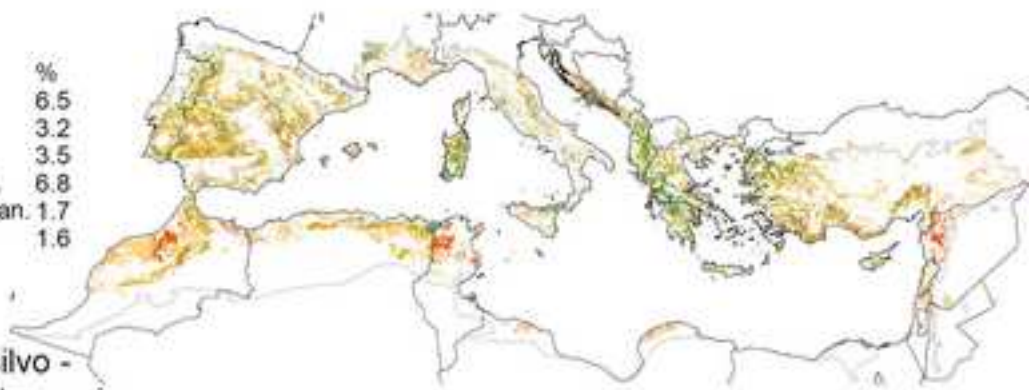
forest systems

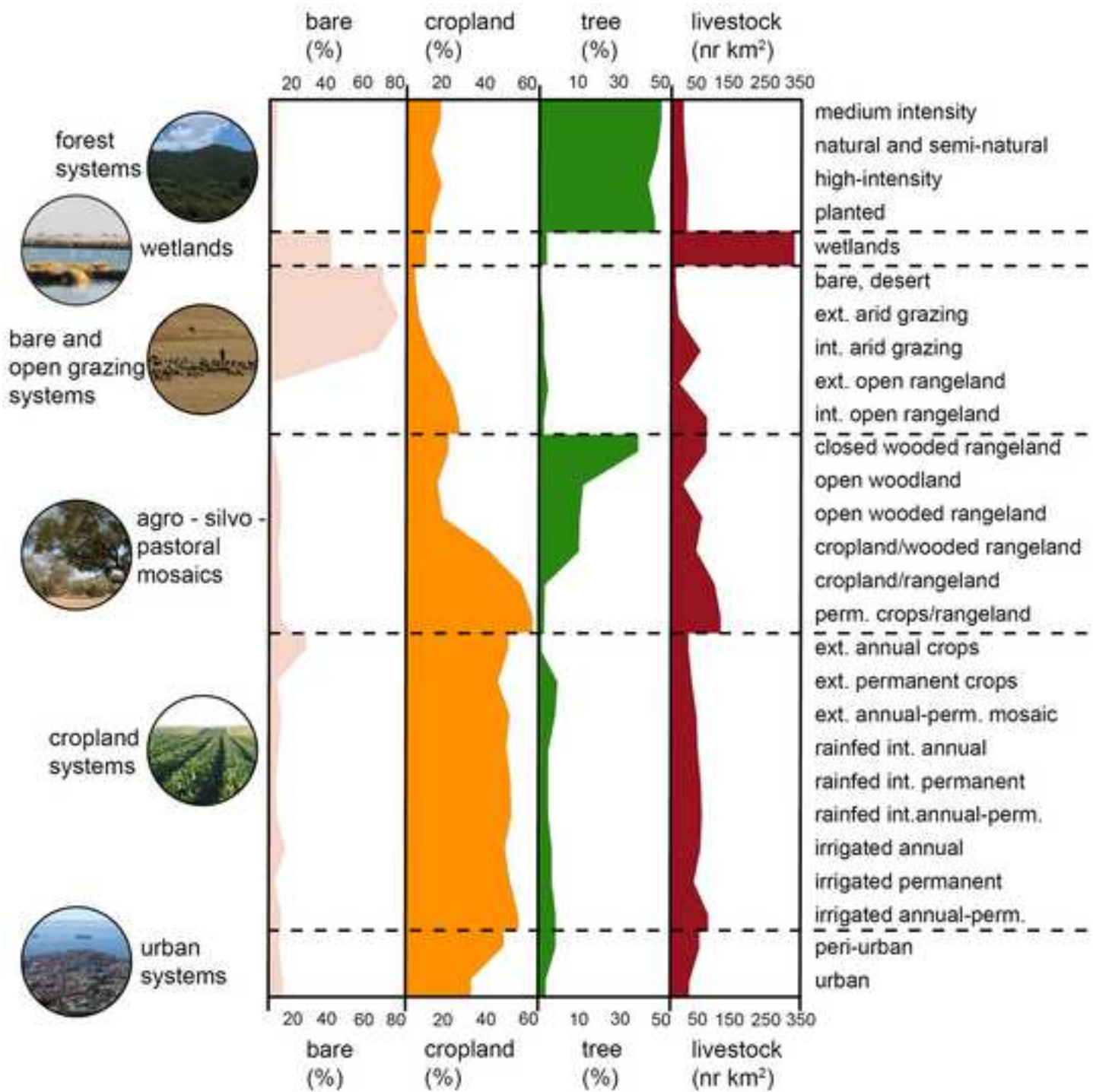


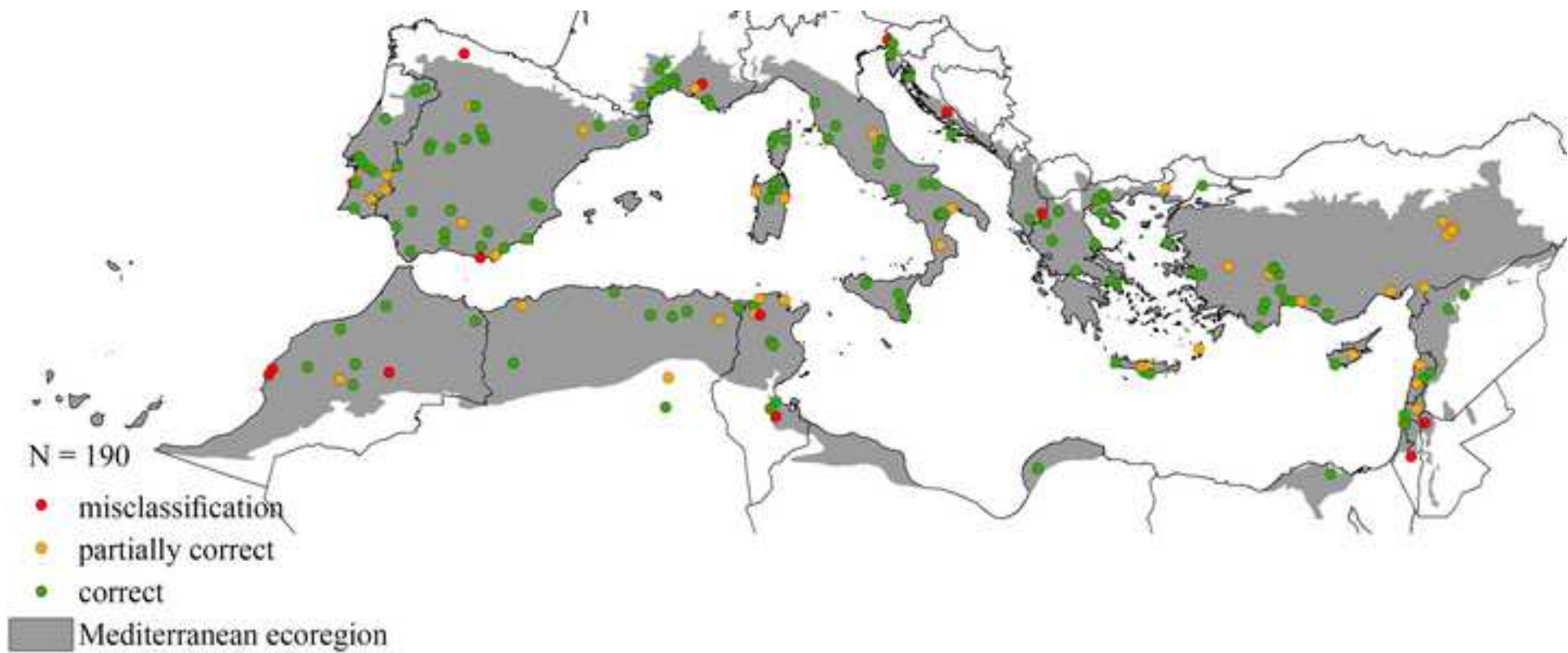
■ cropland/rangeland	6.5
■ open woodland	3.2
■ open wooded ran.	3.5
■ cropland/wooded ran.	6.8
■ perm.crops/wooded ran.	1.7
■ closed wooded ran.	1.6

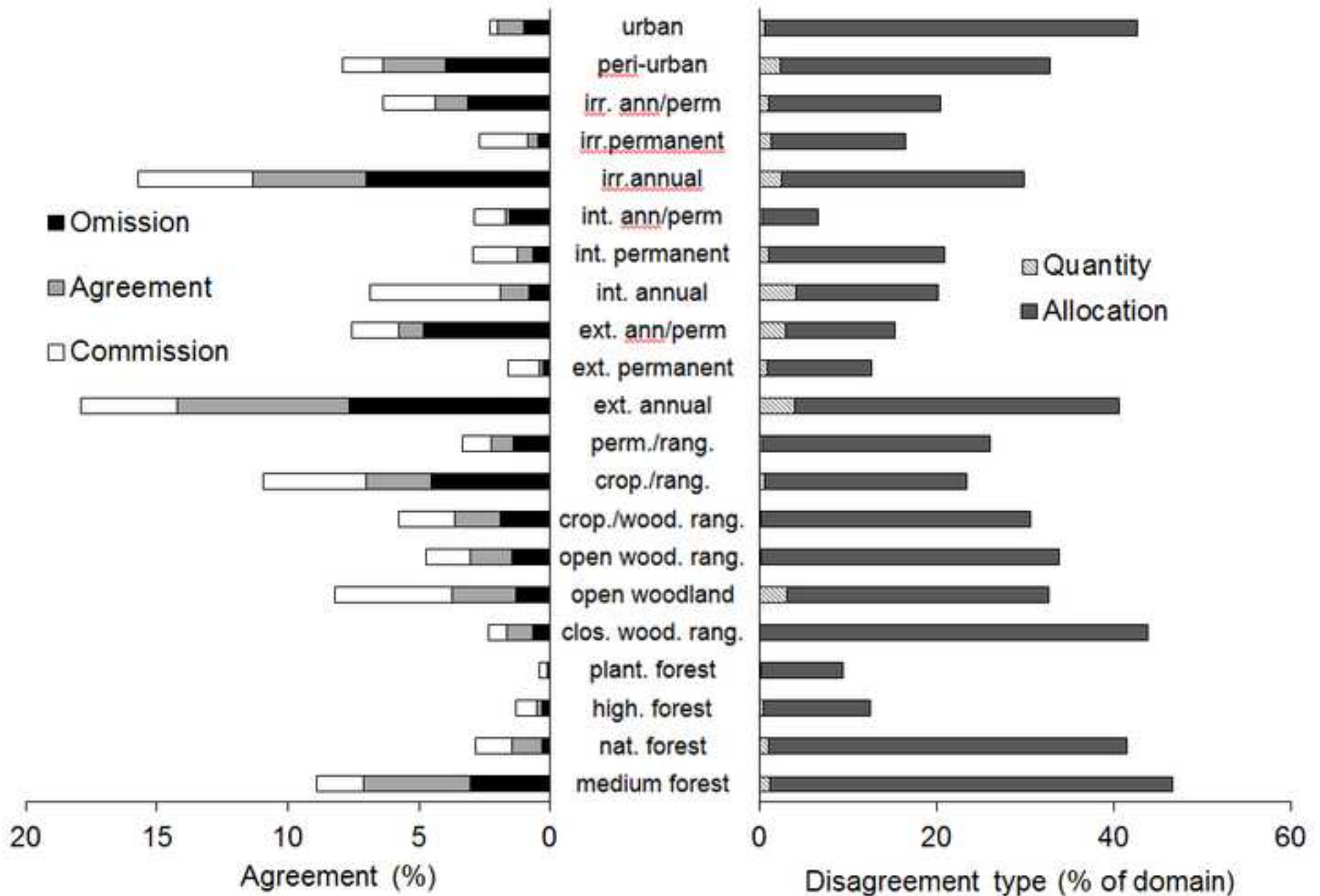


agro - silvo - pastoral mosaics









Acknowledgements

The research in this paper has been supported by the European Research Council under the European Union's Seventh Framework Programme project LUC4C (Grant No. 603542), OPERAs (Grant No. 308393) and ERC grant GLOLAND (no. 311819). We would like to thank Ilse Geijzendorffer, Ana Paula Garcia-Nieto, Marianela Fader, Wolfgang Cramer and Jetse Stoorvogel for their help. This paper contributes to the objectives of the Global Land Project (<http://glp.earth/>).