

1 **A multi-model approach predicts agricultural land abandonment and forest re-**  
2 **growth in an alpine region: a tool for biodiversity conservation**

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9 **Abstract**

10 Land abandonment followed by forest re-growth could have a negative impact on biodiversity by  
11 reducing the species-rich habitats. In previous studies land abandonment is mostly explained by the  
12 recent social and economic development. Particularly agricultural land are abandoned where the  
13 profitability is loss and the cost of management increase. The objective of this study was to predict  
14 land abandonment and biodiversity threat to improve land management and better conservation  
15 practices in an alpine region. We predicted forest re-growth by considering many explanatory  
16 variables comprising climatic, topographic, distance and demographic variables. A multi-model  
17 approach was tested with two different land-cover transition. This permitted us to calibrate the  
18 model on the data set of the first transition and validate it on the data set of second transition. Our  
19 results showed that the model build in the first transition could explain the second transition  
20 indicating that the factors explaining the land abandonment remained constant throughout the time.  
21 We made a projection for the future land abandonment. Additionally we modeled the species  
22 richness over the study areas. Uniting this information with the projected land abandonment  
23 probability we obtained an index of threat to biodiversity loss. Predicting a level of threat for each  
24 agricultural areas gives the possibility to plan some intervention to favour the maintenance of  
25 biodiversity.

26

27 **Keywords:** Agricultural land abandonment, forest re-growth, biodiversity conservation, multi-  
28 model approach.

29

## 30 INTRODUCTION

31  
32 Among the three main threats to biodiversity, i.e. climate changes, habitats destructions and  
33 biological invasions, land use change might be the one where conservation planning could bring the  
34 most notorious improvements. Land use change can be the consequence of intensification of human  
35 activities, but also of land abandonment. Reported changes in land use concern both intensification  
36 of agriculture through fertilization and urbanisation, which drastically reduces the biodiversity or  
37 land abandonment that also have a negative impact on biodiversity. If the intensive use could  
38 represent a threat to biodiversity the effect of agriculture is not always negative. For example in the  
39 case of extensive agriculture by limiting forest recolonization. Recently, large surfaces of  
40 agriculture have been abandoned, particularly in mountain areas, causing a recolonization of the  
41 forest (Gellrich et al., 2007(c), MacDonald et al., 2000).

42 Land abandonment can have some positive impacts, but has mostly a negative impact on  
43 biodiversity. Among the positive consequences, forests permits the stabilization of soil, carbon  
44 sequestration and a temporary increase of biodiversity (Gellrich et al., 2007(c)). However, a main  
45 negative consequence is the long-term loss of open species-rich habitats through competition with  
46 trees (Anthelme et al., 2001). Species richness of all groups (plants, insects) decreases with  
47 increasing cover of trees and shrubs (Öckinger et al., 2006(a)). After abandonment, new species  
48 begin to co-exist with the earlier ones. This cause a temporary increase in species richness and a  
49 negative effect is recorded only after a few years. As many insects groups are depending on the  
50 vegetation, and in particular they are favored by a heterogeneous vegetation structure, they will be  
51 influenced by the cover change. As for the richness of plant species, the abundance and richness of  
52 many animal species increase immediately after an abandonment, but decrease when a forest begins  
53 to establish (Öckinger et al., 2006(a)).

54 The land abandonment observed during the last years in many industrialized countries can  
55 be explained by socio-economic constraints and are useful to understand the increase of the forest  
56 cover occurred in many countries since the mid 19<sup>th</sup> century (MacDonald et al., 2000). The decline  
57 in traditional agricultural practices causing land abandonment has been intensely studied and was  
58 mostly explained by the recent social and economic development. Particularly the loss of  
59 profitability and the increasing cost of agricultural management (Walter 1986; Gellrich et al.,  
60 2007(b)) influenced the abandonment in more remote and less accessible places. As a consequence,  
61 land abandonment occurs primarily in areas where the conditions are less favorable for agricultural  
62 production (higher management cost, more difficult accessibility, lower productivity caused by  
63 unfavorable environmental and geo-physical conditions) (Gellrich et al., 2007(c)).

64 The ability to predict the change from agricultural land to forest land cover can be very  
65 useful to improve land management and better conservation practices. Many studies were  
66 conducted in Switzerland to model these changes (Gellrich et al., 2007(a); Gellrich et al. 2007(c);  
67 Rutherford et al., 2008). These studies were realized by analyzing land-use data from two different  
68 time periods only. With the release this year of a new land-use coverage for the period 2004/09, we  
69 were able to compare land-use changes between three time periods, allowing to analyze two  
70 different transitions. The new second transition permitted to verify if a model built on the transition  
71 between the first and second periods could successfully explain land abandonment and forest re-  
72 growth between the second and third periods.

73 The aim of this study was thus to build a model that could explain and predict the transition  
74 between agricultural patch into forest patch. In particular we wanted know where the transition had  
75 a higher probability to occur. We used geo-physical and socio-economics variables to explain the  
76 forest re-growth in the Alps of the Canton de Vaud and make a predictions of future changes within  
77 the current agricultural surfaces. Additionally we combine the areas with a high level of  
78 biodiversity and a high probability of change in forest to determine the areas with the higher threat  
79 to biodiversity and in need of conservation.

80

- 81 1. We expected that the model made on the first transition could predict accurately and without  
82 loss of predictive power the changes within the second transition.
- 83 2. Second, we expected that the agricultural land where management was costly had a higher  
84 probability to be abandoned and undergo a forest re-growth. We expected to find forest re-  
85 growth principally in areas far from roads and settlement, where agricultural management  
86 costs more energy. Agricultural areas far to forest were expected to have a lower probability  
87 to transit into forest areas in reason of the dispersal capacity of plants (Rickebusch et al.,  
88 2007). Additionally we supposed that forest re-growth occurred in municipality with a  
89 higher emigration and lower economic importance in the primary sector.
- 90 3. Because species richness is higher in extensive agricultural land (that represent habitats with  
91 better quality of nutriments) (Öckinger et al., 2006(b)). the third main hypothesis was that  
92 the land with a higher probability to transit into forest were the areas that had also a higher  
93 biodiversity. We expected to find a concomitance between the areas most supposed to be  
94 abandoned and the areas with a high level of biodiversity.

95

96

97 **METHODS**

98

## 99 **Study area**

100 The study area is situated in the Western Alps of the Canton the Vaud in Western  
101 Switzerland. The area (of ca. 700km<sup>2</sup>) is composed of several valleys ranging from 372 to 3210m at  
102 the top of Diablerets mountains. The land cover is very heterogeneous with several types of open  
103 surfaces, forested surfaces and urban surfaces. Vegetation types are characteristic of the calcareous  
104 Alps, but changes along the altitudinal gradient. The colline belt is covered with deciduous forests  
105 (mainly *Fagus sylvatica*), the montane belt with mixed forests (mainly *Fagus sylvatica*, *Abies alba*);  
106 and the subalpine belt with coniferous forest (*Picea abies*). In the alpine belt, above tree line, the  
107 vegetation consist principally in heath, meadow and grassland vegetation; in the nival belt there is  
108 only a sparse vegetation cover and still important glaciers are found there (Randin et al., 2006).

109

## 110 **Land use transition**

111 The response variable used in the statistical model refers to the occurrence or not of a natural  
112 forest re-growth following agricultural land abandonment. Pixels having transited from open  
113 agriculture area to forest were identified by comparing Land-use data (ASCH) available from the  
114 Swiss Federal Office of Statistics for different time periods (SFOS, 2009). The ASCH is a  
115 nationwide aerial photography aiming at categorizing the land use with a 100m lattice. It is  
116 available for three periods 1979/1985, 1992/1997 and 2004/2009. Among these periods the land use  
117 for our study area were determined at the years 1979, 1992 and 2004. Many studies were conducted  
118 in Switzerland to model the transition of agricultural land to forest land (Gellrich et al., 2007(a);  
119 Gellrich et al., 2007(c); Rutherford et al., 2008) analyzing the GEOSTAT land-use data for the  
120 periods, 1979/85 and 1992/97. Recently the new land-use data available for the period 2004/09  
121 permitted to analyze two different transitions (transition 1 between data of the period 1979/85 and  
122 1992/97, transition 2 between data of the period 1992/97 and 2004/09). In all survey, 72 land cover  
123 classes were distinguished according to the Standard Nomenclature NOAS04 (data from 1979/1985  
124 and 1992/1997 was re-classed in order to be comparables with the new one). For this study we  
125 grouped the land-use classes that interest us in two major groups, “forest” and “agriculture”. The  
126 definition of “forest” and “agriculture” was the same as in the work of Gellrich et al., (2007)(c). The  
127 forest category included: closed forest, open forest, brush forest and wood (classes: 50-60). The  
128 agriculture category include: orchard, vineyard, horticulture, arable land, meadows, farm pastures,  
129 alpine meadows and alpine pasture (classes: 37-49). This category included only the agricultural  
130 classes founded below the tree line. The land-use layers were reclassified into these two categories.

131 The response variables used in the statistical model was a binary variable referred to the  
132 presence or absence of a land-use change. Presence or absence of change was determined by

133 comparing the classification made for the raster layers. Raster's pixels that changed from  
134 "agriculture" to "forest" were labelled as "presence" observations (1), while those that remained  
135 "agriculture" were labelled as "absence" observations (0). We obtained three dataset of response  
136 variables: one for the first transition (1979/85-1992/97), one for the second transition (1992/97-  
137 2004/09) and one for the global transition (1979/85-2004/09). In analyzing the transition 1 and 2 we  
138 determined also which among the agriculture classes had the higher probability to undergo a forest  
139 re-growth.

140

### 141 **Explanatory variables**

142 The explanatory variables were chosen to explain the study hypothesis (Table 1).  
143 Reforestation was supposed to have occurred where the conditions were unfavourable to the  
144 agricultural maintenance. Based on findings from other studies (Gellrich et al., 2007(a); Gellrich et  
145 al., 2007(b); Gellrich et al., 2007(c); Gellrich and Zimmermann, 2007; Rutherford et al., 2008) we  
146 used the variables *Elevation*, *Degree days*, *Precipitation sum*, *Potential radiation* and *Slope* as  
147 climatic and topographic variables available for Switzerland (Zimmermann and Kienast 1999).  
148 These variables were supposed to have an influence on the productivity of the agricultural land. The  
149 decision of abandoning a land can derive, for example, from a lower accessibility to the site, e.g.  
150 caused by too steep slopes.

151 We used *Distance to roads*, *Distance to forest*, *Distance to settlement* and *Distance to*  
152 *agricultural settlement* as distance variables. These variables were used as indicators of  
153 management cost. Agricultural management was considered costly in areas less easily reachable,  
154 i.e. areas far to roads or settlement. The distance to road was calculated as an Euclidian distance  
155 from a digitized network of roads in Switzerland. The distance to forest and settlement (agricultural  
156 and not) was calculated as an Euclidian distance from each corresponding class. The forest category  
157 was the same as for the responses variables, agricultural settlement correspond at class no. 11 from  
158 the NOAS04 Nomenclature, settlement correspond at classes 1-14.

159 Because land abandonment have been linked to rural depopulation and a decrease in farm  
160 employment (MacDonald et al., 2000), we used *Rate of population change*, *Proportion of primary*  
161 *employers* and *Proportion of primary industry* as demographic variables. Data were obtained from  
162 the SFSO and were based on population census undertaken in the years 1968 and 2009. Rate of  
163 population change was calculated as the averaged population change over the twelve years previous  
164 to the determination of land-use for the canton Vaud (1967-79 for period 1, 1980-92 for period 2).  
165 Proportion of primary employers and primary establishment were determined for each commune of  
166 the study area using data from 2001 for the first transition and from 2008 for the second transition.

167 Data on employers and industry correspond at the percent of employers/industry of primary sectors  
168 to the total number of employers/establishment.

169

### 170 **Statistical analysis**

171 We modelled the occurrence of the transition of interest (i.e. agriculture to forest) using the  
172 BIOMOD package (Thuiller et al., 2009) in R (R Development Core Team 2009) 2.10.0.  
173 Explanatory variables were used to model the transition from agricultural area into forested area.  
174 Data of first transition (independent data) were used to calibrate the model, while those of second  
175 transition (dependent data) were used to evaluate it. We used eight different modelling techniques:  
176 Artificial Neural Networks (ANN), Classification Trees Analysis (CAT), Generalised Boosting  
177 Models (GBM), Generalised Additive Models (GAM), Generalised Linear Model (GLM),  
178 Multivariate Adaptive Regression Spline (MARS), Mixture Discriminant Analysis (MDA) and  
179 Breiman and Cutler's random forest for classification and regression (RF). Each of these has  
180 different proprieties and analyses data in a different way. They were all performed at once in  
181 BIOMOD (using the *model* function).

182 For the eight modelling techniques, we compared the importance of the explanatory  
183 variables in explaining the transition between agricultural and forest surfaces.

184

### 185 **Final model and projections**

186 A final ensemble model was computed in BIOMOD. To do this, for each modelling  
187 technique we projected the predicted probability for a transition to occur in the open agricultural  
188 land of the study area. Using the function *Projection*, at each pixel classified as agriculture in 2009  
189 was given a probability of making a transition during the future years. As each modelling technique  
190 possesses its own strengths and weaknesses, we used the Ensemble modelling approach to combine  
191 the single projections of each models to obtain a single values for each pixel (Araujo and New,  
192 2007). Probability values range from 0 (estimated probability of 0% to transit into forest) to 1  
193 (estimated probability of 100% to transit into forest). We obtained a final map that displays the  
194 probability of each pixel classified as agricultural surface in the ASCH 2009 to transform into  
195 forest.

196

### 197 **Biodiversity modelling and assessment**

198 During the years 2002 to 2009, 912 vegetation plots were sampled throughout the study  
199 area. The diversity of species has been modelled across the entire study area using a generalized  
200 linear model (GLM) with a Poisson distribution. For each pixel of the study area, the model

201 predicted the diversity of species. The explanatory variables taken into consideration were: *Degree-*  
202 *days*, *Moisture index* (average of values of June-August months), *Solar radiation*, *Topographic*  
203 *position* and *Slope*. We then transformed species richness in a number between 0 and 1 (1  
204 corresponded at the pixel with the highest biodiversity value) We done this transformation in order  
205 to have a value easily comparable to the value of probability transition.

206 For the agricultural pixels correspondent at the vegetations plot we extract the probability to  
207 undergo a change calculate with the final model. We than test if a correlation existed between the  
208 two variables. In order to give at each pixel in the study area an index of its condition (at risk vs.  
209 non-at risk) we combined the information about land-use transition with those about biodiversity.  
210 We multiplied the transition probability and the biodiversity level to obtain a value between 0 and  
211 1. Values close to 0 correspond to areas with a low biodiversity and a low probability to undergo a  
212 transition to forest. Values tending to one correspond to areas with a high biodiversity level and a  
213 high probability to undergo a transition to forest.

214

215

## 216 **RESULTS**

217

### 218 **Model performance**

219 Our model successfully predicted forest re-growth for transition 1 (dependent sample) and  
220 could also largely explain forest re-growth occurring in the second transition (independent data).  
221 Prediction accuracy measured by the area under the ROC curve (AUC) measured on the dependent  
222 data set ranged from 0.70 to 0.79. The same AUC measured for the independent data set (range:  
223 0.61-0.76) were almost always slightly lower than those measured on the calibration data (Table 1).  
224 According to De Leo (1993), AUC value can be translated in the probability that the model could  
225 correctly explain the transition. This means that our model could distinguish between presence and  
226 absence of forest re-growth with a probability of 0.76% and 0.75% (mean of the respective AUC  
227 range) (Figure 1). These values indicated that the model was useful to make a projection.

228

### 229 **Explanatory variables importance**

230 The different explanatory variables had varying importance in the model (Figure 2).  
231 Modelling techniques analysed the data in different ways, it is why the importance value could have  
232 a great variance. Here we chose to consider the importance mean as an indicator of the importance  
233 of the variables in explaining the forest re-growth. The importance mean corresponds to the mean of  
234 the value that the variable of interest took across models fitted with the different modelling

235 techniques. *Degree day* is the only one with a importance mean greater than 0.5. *Elevation*  
 236 (mean=0.32) and *Distance to forest* (mean= 0.33) were also important in the model, and to a lesser  
 237 extent *Slope* (mean=0.17) and *Precipitation* (mean=0.15). *Radiation*, *distance to settlement*,  
 238 *Distance to agricultural settlement*, *Distance to roads*, *Demographic change*, *Employers I* and  
 239 *Industry I* showed only very little or no significant relationship to the response variable.

240 Looking at graphs of the variable response curves, *Elevation* shows a positively relationship,  
 241 suggesting that forest re-growth occurs principally at high elevations. *Distance to forest* shows, as  
 242 expected, a negative relationship suggesting that forest re-growth occurs only in areas far from  
 243 forests. The most important variable, *Degree day*, shows a positive relationship, suggesting that  
 244 forest re-growth has a major probability to occur in areas with a value greater than 2000 KJ/day.  
 245 *Slope* and *Precipitation* both show a similar trend, with a greater probability of forest re-growth in  
 246 areas with intermediate values of slope and precipitation (an average between 30°-60° and 1400-  
 247 1900 [1/10mm/month]). As *Precipitation* is correlated with elevation, a similar trend was found for  
 248 these two variables. Although the other variables did not show any important relationship to the  
 249 response variable, we noticed a slight tendency for forest re-growth to occur in areas far to  
 250 agricultural settlements. Demographic variables did not show any importance in any modelling  
 251 technique.

252

### 253 **Projection**

254 The representativity of the model permitted us to make a prediction for the future years.  
 255 With the 'Projection' function, we obtained a probability of change for each pixel classified as  
 256 agricultural land in 2009 (Figure 3). The probability obtained refers to a possible transition in the  
 257 futures years and thus the resulting maps reveal those agricultural areas most exposed to suffer a  
 258 transition to forest. These are principally located in the middle of large forests.

259 In analysing the transitions occurring between 1979/85 – 1992/97 and 1992/97 – 2004/09, we  
 260 determined which agricultural classes had undergone a transition to forest. We found that a change  
 261 was observed mainly in the classes "Favourable alpine pastures"(class 46) and "Brush alpine  
 262 pastures"(class 47). This permitted us to additionally predict that agricultural areas with low direct  
 263 use will have the highest probabilities of conversion to forest (Annex 1). The analysis of transition  
 264 also permitted to note that most pixels that underwent a transition had changed to "open forest (on  
 265 agricultural areas)" (class 55) and to "Clusters of tree (on agricultural areas)" (class 59). We remark  
 266 that they were mainly forest surfaces developed on formerly agricultural surfaces, which confirms  
 267 that the forest re-growth happened generally to the detriment of agricultural land (Annex 2). This  
 268 observation confirmed previous findings by Rutherford and co-workers (2008).

269

**270 Conservation of Biodiversity**

271 Biodiversity level and transition probability were positively correlated ( $\text{cor} = 0.259$ ,  $p <$   
272  $0.001$ ), (Figure 4). This confirmed the hypothesis that the areas with a higher probability to be  
273 abandoned are also the areas with a high level of biodiversity. The biodiversity map showed that  
274 species richness was higher principally in mid altitude areas near to forest. Particularly endangered  
275 areas - i.e. those with high biodiversity level and a high probability of transition to forest – may  
276 correspond to land loosing a great part of its current species richness. We founded that these areas  
277 were mainly localized in the middle of larges forest surfaces at intermediate elevations. They were  
278 rarely found at higher or lower elevation (Figure 5).

279

280

**281 DISCUSSION**

282

283 Forest re-growth is an important threat to biodiversity. Its increase cause the disappearance  
284 of many grassland species and animals. Several studies (Gellrich et al., 2007(a); Gellrich et al.,  
285 2007(b); Gellrich et al., 2007(c); Gellrich and Zimmermann, 2007; Rutherford et al. 2008) had  
286 modelled the increase of forest linked to agricultural land abandonment to better explain the  
287 phenomenon. The results obtained showed that agricultural land abandonment cold be explaining in  
288 considering some climatic, topographic, economic and demographic variables among a period of  
289 time. In our study we founded that the phenomenon of forest increase could be modelled but also  
290 predicted by a model. This is a very important point that permit to better understand and monitoring  
291 this phenomenon.

292 More in particular the results of our model showed that the mean of AUC values were  
293 similar for the dependent data (0.76) and the independent data (0.75). These values did not  
294 correspond to a very high explanatory power, but they were better than the AUC values found in  
295 other studies, i.e Gellrich et al., 2007(c). Despite, these difference values obtained are not totally  
296 comparables because of the different approach used in this two study (different model, different  
297 study area size). Anyway the values obtained from the two data sets were very similar and quite  
298 high, so that the results obtained confirmed our first hypothesis. The model calibrated with the first  
299 transition could be applied to the second transition data keeping a useful AUC. This suggested that  
300 the factors explaining the first transition remain quite stables over the years. This stability is an  
301 important factor that permitted us to make a model over the total transition, apply it to the present  
302 and make a projection for the futures years. The possibility to make a projection is an important tool

303 to predict the future development of the forest re-growth. The multi model approach used in this  
304 study is a recent methods that permit to better analyse the data. In particular the niche based model  
305 permitted to analyse the data with several methods and so have a stronger approach using the  
306 Ensemble. Making an ensemble permitted to make a better projection of the probability to a forest  
307 re-growth. However, there were some important factors explaining the land cover change that could  
308 not be considered in this study, such the individual motivation of a land abandon and the local  
309 characteristics of land.

310 The second hypothesis was partially confirmed by the results that suggested that forest re-  
311 growth took place where the cultivation costs were high and yield potential was low. Some of the  
312 explanatory variables showed a high importance in the model and the results gave precise indication  
313 about the occurrence of forest re-growth. This occurrence was found principally in favourable  
314 alpine pastures, brush alpine pastures or in areas at mid to high altitudes with a higher temperature  
315 sum, mid- steep slopes, mid-high level of precipitation. Most of the overgrown areas have been  
316 found in areas closer to the forest and, with a slightly effect, farther to agricultural settlement.

317 Variable elevation showed a positive relationship with forest re-growth indicating that the  
318 transitions were most frequent on high altitudes. This agree with the expectation that surfaces in the  
319 bottom of valley were more exploited because more easily accessible and economically exploited  
320 (Gellrich et al., 2007(b)). Interesting was the contraposition between the influence of the altitude  
321 and those of the temperature. The variable degree days showed in fact a positive relationship with  
322 the frequency of forest re-growth, indicating that a forest re-growth occurred at high altitudes (not  
323 too high altitude), but only where the temperatures were sufficiently high too permit the  
324 development of forest (Gehrig-Fasel et al., 2007).

325 Forest re-growth occurred more frequently on land with intermediate measures of steepness.  
326 Slightly and steep measures of slope undergone a transition less frequently. Probably this is due to  
327 the different characteristic of these surfaces. Slightly slopes are generally located on the bottom of  
328 valley these surfaces are intensively exploited for agricultural production, land cultivation is  
329 maintained due to the high suitability, an abandon is very improbable. Mid-steep slope are generally  
330 located at mid-higher altitudes and used for pasturing (Netting, 1972). This land use is related to  
331 low cost but, in case of abandon, these surfaces could change into forest land. Steeper slope are  
332 located at higher altitudes near the tree line, if abandoned the land forest re-growth is limited  
333 because of unfavourable growing conditions for trees causing by the high elevations, the lower  
334 water holding capacity and the snow avalanches (Gellrich et al., 2007(a),(c)). When we chose the  
335 agricultural categories to analyze we tried to have only grassland areas below the tree line. But, the  
336 past intensive cutting had down-shift this limit so the position of the tree line is unclear and we had

337 to consider all the subalpine surface. This can explain why the surfaces abandoned were principally  
338 the “Favour alpine pastures” and why at steeper slope the frequencies of forest re-growth is  
339 reduced.

340 The variable distance to forest was significant in the model and showed the expected trend.  
341 Forest re-growth decreases with increasing distance to forest. The negative relationship can be  
342 explained in consider the dispersal capacity of the plants (Rickebusch et al., 2007). Land far to  
343 forest will be more difficultly and slowly colonised by plant. Another explanation is more related at  
344 the decision of abandonment. We can suppose that agricultural areas nearer to forest are also the  
345 areas with smaller size. It is costly manage a great number of small and separated areas than a few  
346 but higher areas (Gellrich et al., 2007(a),(c)). When a farmer want reduces his territories is more  
347 probably that he will abandon first the smaller areas. This mechanism is know as gap filling.

348 Distance to road was not significant in the model. The effective influence of this variable is  
349 still unclear. Some studies found that distance to road was negatively correlated with the forest re-  
350 growth (Pezzati, 2000). Oppositely, other found that the land use decrease with the increasing  
351 distance to roads (Gellrich et al., 2007(c)). These variations were supposed to be due at the different  
352 use of these surfaces. On remote mechanically exploited meadow the harder accessibility with  
353 agricultural vehicles can cause the abandon of the land. On pastures land the remoteness has minor  
354 importance and the abandon is less probable. This means that the importance of this variable in the  
355 model can vary depending of the frequency of these two different land-uses in the area considered.  
356 Another aspect determining the influence of this variable on the model is the precision of the road  
357 network used. The routs map was too precisely and sometimes contained also very small roads and  
358 impassable roads that they could influence the result.

359 Surprisingly, we found no influence in forest re-growth of the proportion of industry and  
360 employees in the primary sector. This could be due at the origin of data used in the model.  
361 Demographic data from 2001 used to explain the transition were not sufficient pertinent and  
362 precise. However the number of industry was not closely related to the level of agricultural land  
363 exploitation. Recent innovations and technologies permit to exploit large areas employing few  
364 workers so that it was not possible explain the agricultural abandonment using the proportion of  
365 primary industry and primary employers in the municipalities.

366 Rate of population change was also not relevant in the model. The variation in population  
367 change (increase or decrease) between the last 20 years did not show a precise trend and probably  
368 they could not have a strong influence on land abandonment phenomenon. Additionally the limited  
369 extension of the study area did not permit to have a strong association between the number of  
370 habitant par commune and the rate of forest re-growth. Rate of population change was an indicator

371 of emigration/immigration but the value was not corrected with the municipality surface or the total  
372 number of habitants. A municipality could show a decrease of population in keeping a high  
373 proportion of people in confront of the area surface. The others study showed also any relation  
374 between emigration and forest increase (Gellrich et al., 2007(d)), although the rural depopulation  
375 was consider as the cause of land abandonment in many European mountain regions (MacDonald et  
376 al., 2000).

377 Distance to settlement was insignificant in the model, while distance to agricultural  
378 settlement was slightly important and had an influence according with our initials expectations. This  
379 suggests that to consider all the construction was probably less informative and could not explain  
380 the transition. However consider only the distance to agricultural settlement had a higher influence  
381 despite it remain quite insignificant. Higher distances to settlement represent a higher cost of  
382 management but, as suggest for the influence of demographic variables, the technology could have  
383 minimized this negative effect.

384 Because the forest re-growth had a negative impact on biodiversity, predicting the land  
385 abandonment could be very useful and important to make a conservation planning. An increasing  
386 cover of tree and shrubs decrease the species richness of all groups in the concerned areas. Species  
387 richness of many groups such plants, birds and insects are favoured by an heterogeneous vegetation,  
388 so many times it first increase when management and grazing are stopped. However, as the  
389 reforestation continues the number of grassland species strongly decrease (Öckinger et al., 2006(a)).  
390 The biodiversity conservation is a main theme among the ecological biology. The maintenance of a  
391 high specific richness is the purpose of many studies. In this study we used the results of our model  
392 as a possible point of start of a conservation planning. Because our model could predict the  
393 transition from agricultural land to forest land, we combined the projection for the future land cover  
394 transitions with an index of biodiversity to highlight the areas with a higher threat of biodiversity  
395 loss. As we predicted the result showed that the most threat areas were located in correspondence of  
396 the areas most subject to undergo a transition. A logical explanation exists for the positive  
397 relationship between the probability of a transition and the level of biodiversity. A first step of an  
398 abandonment generally occurs when a land intensively used is transformed in an extensively-used  
399 surface. An extensively used, such a pasture land, has less or no economic investment so that this  
400 transition could be an alternative of a total abandonment. Therefore, total abandon occurs  
401 principally in extensively used land. These areas have a higher probability to be abandoned but they  
402 have also a higher level of biodiversity in reason of their less-invasive use (Öckinger et al.,  
403 2006(b)).

404 Predicting a level of threat for each pixel gives the possibility to plan some intervention to favour  
405 the maintenance of biodiversity. In particular the map of threat level permits to determine the areas  
406 with the most urgent need of protection. In function of the resources at disposal we could determine  
407 an adequate number of areas that could be subject of a conservation planning.

408 In conclusion our results represent a tool to predict the loss of biodiversity and the species range  
409 shift on open areas. But the forest increase and the reduced habitats of mountain grassland species  
410 are not only the consequence of the agricultural land abandonment. Other factors as the progressive  
411 climate warming is shifting the species range on open areas and the effect of climate is often used to  
412 model the plants distribution (Randin et al., 2009; Engler et al., 2009). An approach including both  
413 land abandonment and climate change could be useful to better understand the land cover change  
414 and its effect on the species richness and distribution.

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416

#### 417 **Acknowledgments**

418

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420 Office of Statistic for supplying data and Federico Tettamanti for his help to start this work.

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**Table-1 Potential explanatory variables**

Variable names	Unit	Exp. Sign <sup>a</sup>
<b>Climatic and topographic variables</b>		
Elevation (E)	m	+
Degree days (DD)	(day °C)	+
Slope (S)	(°)	+
Radiation (R)	(KJ/day)	?
Precipitations (years) (PR)	(1/10mm/mounth)?	
<b>Distance variables</b>		
Distance to forest (d_F)	(m)	-
Distance to settlements (d_S)	(m)	+
Distance to agricultural settle (d_SA)	(m)	+
Distance to road (d_R)	(m)	+
<b>Demographic variables</b>		
Demographic change (POP)	(mean/10 years)	-
Industry I (I_I)	(%)	-
Employers I (E_I)	(%)	-

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<sup>a</sup> the expected sign represent the influence that we expected to find according to other studies.

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499 **Table 2 – AUC of each modelling methods with an**  
500 **Evaluation Roc**

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	<b>Calibration data <sup>a</sup></b>	<b>Validation data <sup>b</sup></b>
501 <b>ANN</b>	0.703	0.605
502 <b>CTA</b>	0.745	0.738
<b>GAM</b>	0.784	0.757
503 <b>GBM</b>	0.785	0.760
<b>GLM</b>	0.787	0.753
504 <b>MARS</b>	0.780	0.736
505 <b>MDA</b>	0.752	0.699
<b>RF</b>	0.776	0.743

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506 **a** calibration made with variable response from  
507 transition1 (1979/85-1992/97).

508 **b** validation made with variable response from  
509 transition2 (1992/97-2004/09).

510 **LEGEND**

511

512 **Figure 1:** Boxplot of AUC of the model. The calibration model (data from transition1) shows an  
513 accuracy above 0.7. The validation model has an accuracy range between 0.61 and 0.76.

514

515 **Figure 2:** Importance of the explanatory variables in the models. The climatic and topographic  
516 variables have the greater importance, particularly elevation, degree day slope and precipitation  
517 contribute in main part to explain the transition. The distance variables contributed with the distance  
518 to forest and in minor way with the distance to agricultural settlement. The demographic variables  
519 don't contribute at the model.

520

521 **Figure 3:** Representation of the probability to undergo a change into forest surface for each pixel of  
522 the study area classified as agricultural in 2009.

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524 **Figure 4:** Probability of a change reported in function of the level of biodiversity. In red is reported  
525 the linear regression. The two variables are positively correlated.

526

527 **Figure 5:** Representation of the level of threat to biodiversity of the study areas. In red there are  
528 areas with a high value of threat that correspond at a high value of biodiversity and a high  
529 probability to undergo a change into forest land. In blue there are areas with low value of threat.

530 **Figure 1.**

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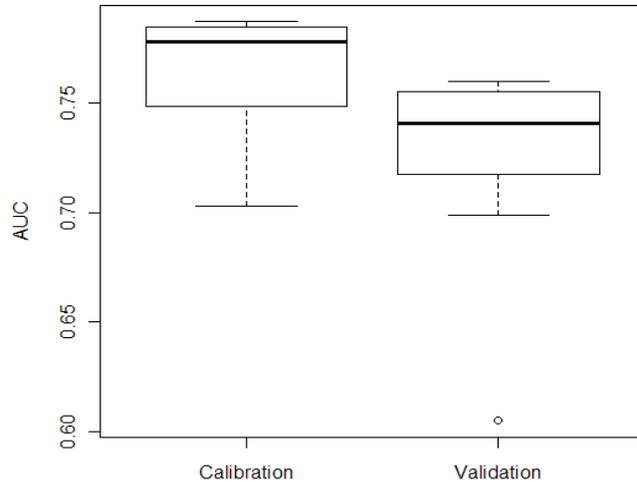
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543 **Figure 2.**

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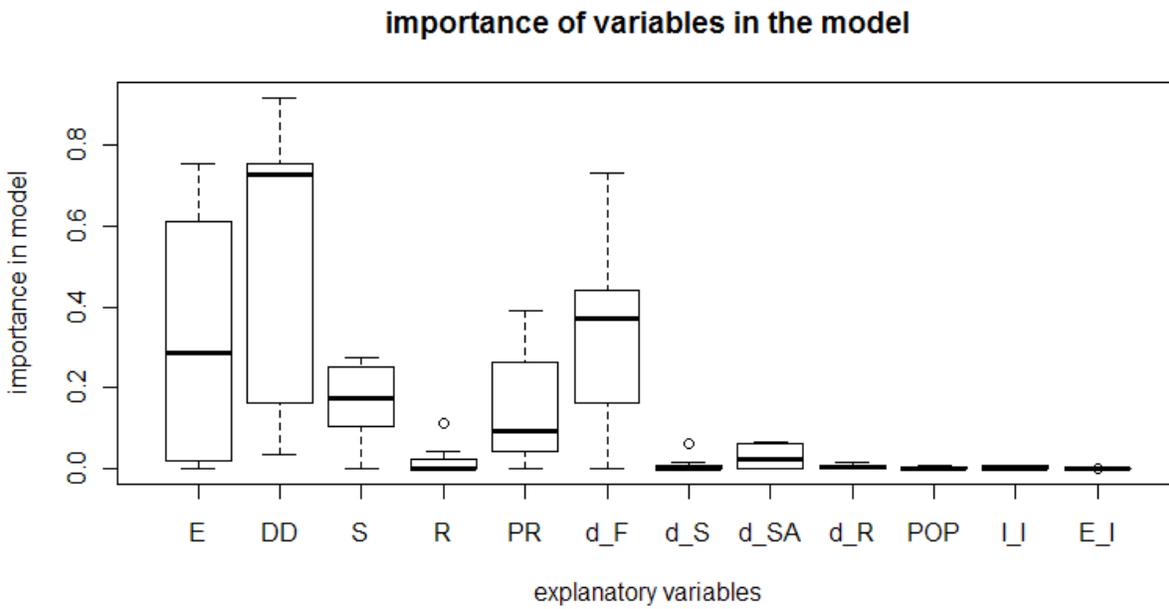
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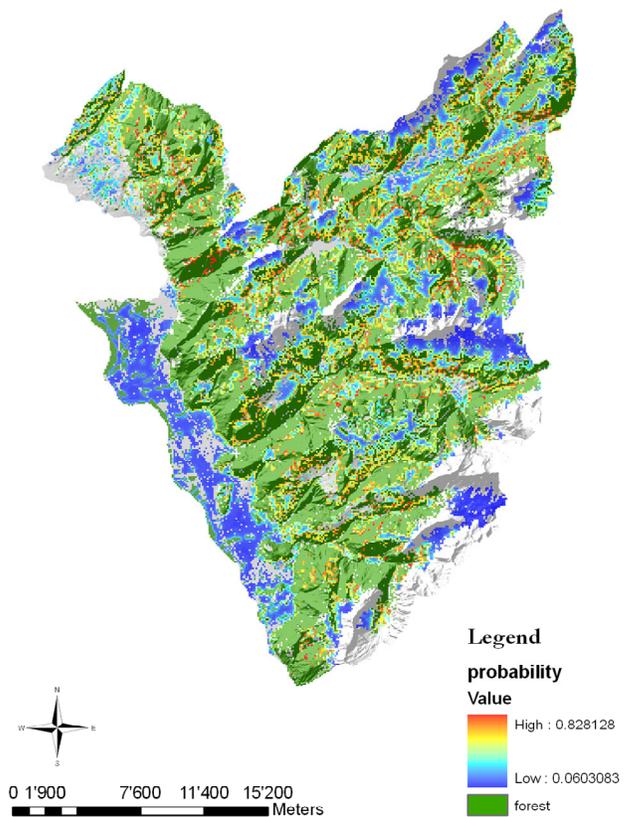
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557 **Figure 3.**

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578 **Figure 4.**

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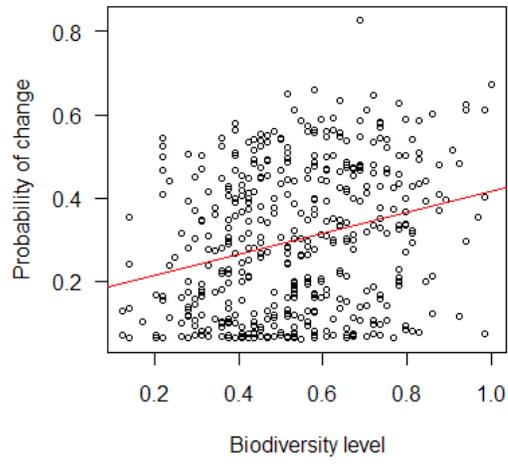
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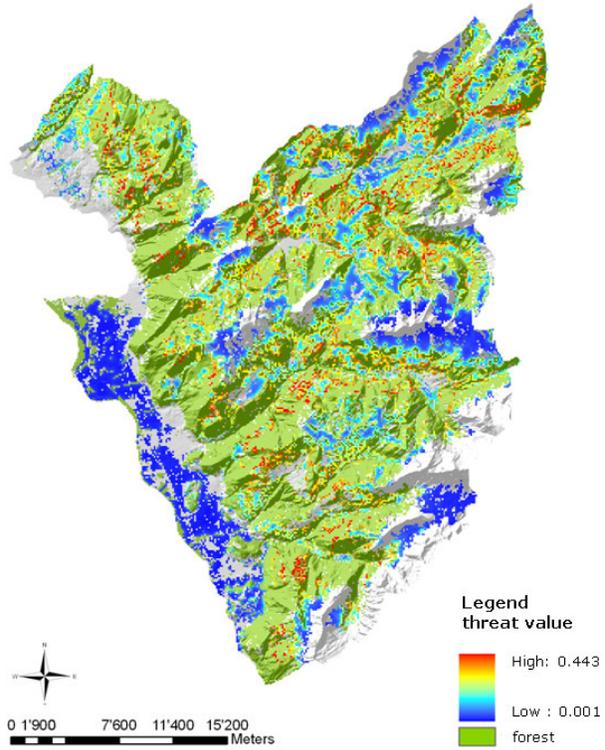
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593 **Figure 5.**

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613 **Annex 1.**

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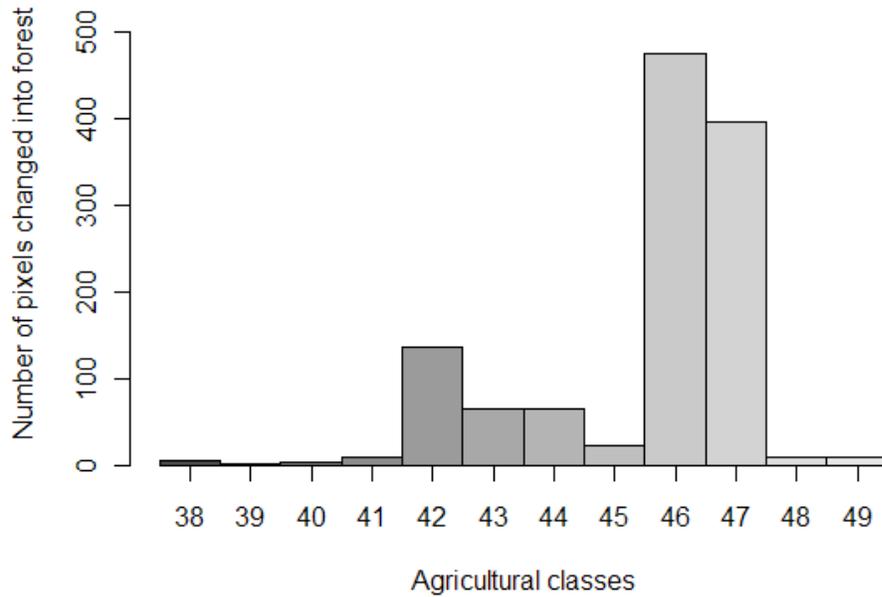
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**Annex 1:** Total number of agricultural pixels that had undergo a transition into forest from 1979 to 2009. The x axis represent the agricultural classes numbered in accord with the NOAS04. The classes 46 and 47 correspondent at “Favourable alpine pastures” and “Brush alpine pasture”. A transitions occurred mostly in these to classes.

634 **Annex 2.**

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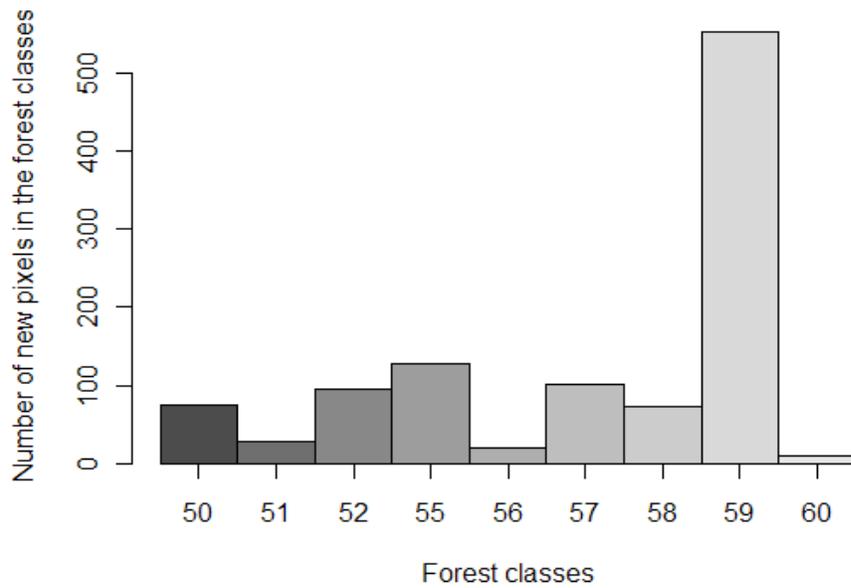
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**Annex 2:** Total number of new forest pixels that had recorded from 1979 to 2009. The x axis represent the forest classes numbered in accord with the NOAS04. Agricultural pixels that undergone a change became mostly classified as “Cluster of trees (on agricultural areas) that correspond at the class 59.

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