

LecoS - A QGIS plugin for automated landscape ecology analysis

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Abstract:

The quantification of landscape structures is an important part in many ecological analysis dealing with GIS derived satellite data. This paper introduces a new free and open-source tool for conducting landscape ecology analysis. LecoS is able to compute a variety of basic and advanced landscape metrics in an automatized way by iterating through an optional provided vector layer. It is integrated into the QGIS processing framework and can thus be used as a stand-alone tool or within bigger complex models. Finally a potential case-study is demonstrated, which tries to quantify pollinators responses on landscape derived metrics at various scales.

Key-words: QGIS, automation, landscape ecology, landscape metrics, Python, GIS tools, pollinators

Introduction:

The use of free and open-source software in ecological research has gained increasing attention in the last years (Steiniger & Hay, 2009; Boyd & Foody, 2011). Freely available open-source software has several advantages in research such as that the computational and statistical background of the analysis can be independently investigated and verified. Furthermore free software can enhance biological research and knowledge transfer in developing countries, where financial constraints can prevent the access to proprietary alternatives (Steiniger & Hay, 2009).

Within ecological research the field of landscape ecology features a number of free and open-source tools (Steiniger & Hay, 2009). Scientific studies in landscape ecology study the relationship between spatial patterns and ecological processes on a variety of spatial and organizational levels (Turner, 1989; Wu, 2006). Landscapes are here often seen as mosaics of differently structured and composed land-cover patches which are potentially connected by spatial dynamics (Pickett & Cadenasso, 1995). The landscape structure can be quantified by size, shape, configuration, number and position of land use patches within a landscape. Those quantified values and metrics are invaluable for various fields of ecological research like for instance studies on the influence of habitat fragmentation on wildlife (Fahrig 2003).

Landscape metrics are usually derived from classified land-cover datasets using specialist software and graphical information systems (GIS). See Steiniger & Hay (2009) for an extensive overview of freely available open-source software for landscape ecologists. Out of those software products FRAGSTAT is most likely the most comprehensible software package for the calculation

40 of landscape and patch metrics (McGarigal & Marks, 1995; McGarigal et al., 2012). However the
 41 analysis in FRAGSTAT is separated from the visualization in a GIS program and does not run
 42 natively on all operating systems such as Mac-OS or Linux derivatives. Other widely used
 43 open-source software suites include the r.li extension for GRASS GIS (Baker & Cai, 1992) and
 44 SDMTools for the R software suite (VanDerWal et al., 2012). Those solution however depend on
 45 prior raster formating and cropping or can not be used in complex hierarchical models without
 46 knowledge of programming or scripting.

47 Here a new tool is introduced which is capable of analyzing various landscape and patch
 48 metrics within a freely available open-source GIS suite and is thus being able to combine the ability
 49 of calculating complex landscape metrics within sophisticated GIS models.

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 51 **Landscape ecology analysis in QGIS**

52 The QGIS project provides a free and open source desktop and server environment and ships
 53 with all functionalities of a modern GIS system (QGIS Development Team, 2013). It furthermore
 54 allows the easy extension of its core functions through user-written plugins, which can be
 55 downloaded within the desktop suite. Since the last stable version – codename 'Dufour' – the
 56 popular spatial data processing framework SEXTANTE has been integrated into QGIS. This new
 57 'Processing toolbox' not only integrates existing geoprocessing functions into a similar toolbox as in
 58 the prominent ArcGIS suite, it also allows the creation of automatized models, which are able to
 59 combine several individual spatial calculations into single sequential models. Additionally, users are
 60 able to add their own python or R scripts to the Processing toolbox.

61 Here a new plugin for QGIS called LecoS (**Landscape ecology Statistics**) is introduced. It
 62 makes heavy use of the scientific python libraries SciPy and Numpy (Jones et al., 2001; Oliphant,
 63 2007) to calculate basic and advanced landscape metrics and provides several functions to conduct
 64 landscape analysis. Up to now over 16 different landscape metrics are supported. LecoS
 65 furthermore comes with two different interfaces. Core functions like the computation of landscape
 66 metrics have their own graphical interface, while more advanced functionalities are only supported
 67 in the QGIS Processing toolbox.

*Table 1: List of functions to date (Version 1.9.2). All functions need installed **python-osgeo**, **python-scipy** and **python-pil** bindings within QGIS 2.0.1 Dafour.*

Name	Interface (Graphical Processing)	Description
Landscape preparation		
Create random landscape (Distribution)	no yes	Allows to create a new raster layer based on a chosen statistical distribution. The user can specify the

		extent of the output and distribution parameters.
Intersect Landscapes	no yes	Takes a source and target raster layer as input and calculates the intersection of both layers.
Match two landscapes	no yes	Reprojects and interpolates a raster layer to the projection and extent of a target raster.
Landscape statistics		
Count Raster Cells	no yes	Returns the number of cells per unique cell value inside a raster layer
Landscape wide statistics	yes yes	Allows to calculate various landscape metrics for an input raster layer
Patch statistics	no yes	Computes patch metrics for a given land cover class.
Zonal statistics	no yes	Performs a zonal statistics analysis with a raster layer containing zones and a raster layer containing values as input.
Landscape vector overlay		
Overlay raster metrics (Polygons)	yes yes	Allows to compute landscape or patch metrics for each polygon feature of an input vector layer. Results can be generated as new separate table or added to attribute table of the vector layer.
Overlay vector metrics (Polygons)	yes no	Can calculate basic metrics for attribute derived classes inside a polygon vector layer.
Query raster values (Points)	no yes	Returns all raster values of the cells below a given point layer
Landscape modifications		
Clean small Pixels in patches	yes yes	Cleans a given classified raster layer of small isolated pixels.
Close holes in patches	yes yes	Closes holes (inner rings) in all patches of a specified land cover class.
Extract patch edges	yes yes	Extracts the edges from each patch of a given land cover class.
Increase/Decrease patches	yes yes	Allows the user to increase or decrease all landscape patches of a given land cover class.
Isolate smallest/greatest patches	yes yes	Returns a raster layer with the greatest or smallest identified land cover patch. If multiple patches fulfill this criteria, than all of them are returned.
Label Landscape patches	no yes	Conducts a connected component labeling (chessboard structure) of all raster cells with a given value. The output contains a raster layer where all individual patches have a single unique identifier.

Neighbourhood Analysis (Moving Window)	no yes	Calculates statistics for cells in a raster layer using a moving window approach.
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69 Since LecoS version 1.9 the set of available functions can be divided into the categories
70 *Landscape preparation*, *Landscape modification*, *Landscape statistics* and *Landscape vector*
71 *overlay* (Table 1). Landscape preparation functions allow the user to prepare and match input layers
72 to each other, while landscape modification functions can modify or generate derivatives of raster
73 layers. Users can calculate landscape metrics or raster properties with the Landscape statistics
74 functions and are also able to automatize those calculations for all features of a given vector layer
75 (Figure 1).

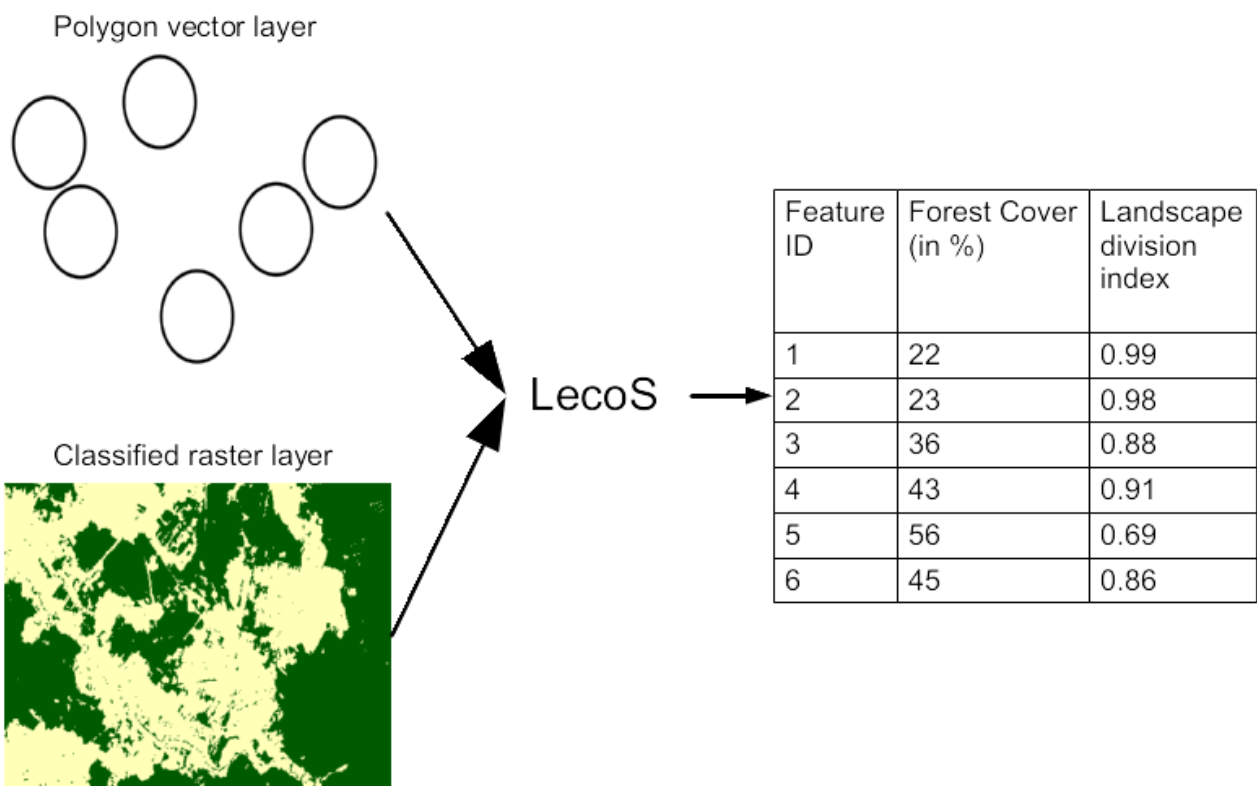


Figure 1: Illustrating the power of the Landscape vector overlay functions. The intended goal is to calculate the percentual proportion of forest cover and Jaegers landscape division index for every single study site (Jaeger, 2000) Using the vector overlay function LecoS is able to automatically compute the selected landscape metrics for every feature of the provided vector layer.

77 LecoS can be acquired through the QGIS plugin manager or directly downloaded from the
78 QGIS plugin hub (<http://plugins.qgis.org/plugins/LecoS/>). The python libraries SciPy, NumPy and
79 the imaging library PIL have to be installed and correctly configured in QGIS beforehand.

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83 **Case-study**

84 In the following paragraphs a exemplary use-case of LecoS is demonstrated using real
85 sampled field data. Pollinators are of increasing concern in recent ecological studies due to their
86 reported declines which might endanger ecosystem service provisioning worldwide (Klein et al.,
87 2007; Potts et al., 2010). The drivers of pollinator decline are various and habitat loss is certainly
88 one of many reasons (Potts et al., 2010). Semi-natural habitats provide resources and nesting
89 opportunities for many pollinators such as bees or hoverflies. However not only the primary habitat,
90 but also the surrounding landscape is important as resource. Here it has been shown that different
91 taxonomic units of pollinators show opposing responses to landscape structure and remoteness from
92 their primary habitat (Meyer et al., 2009; Jauker et al., 2009). The goal of this little case-study will
93 be to characterize the heterogeneity and characteristics of the landscape at various scales using an
94 automatized model framework within QGIS.

95 All locality and species data was taken from Mudri-Stojnić et al, (2012), who conducted an
96 assessment of pollinator abundance and diversity in 16 grassland habitats in northern Serbia. The
97 study sites were all located within or next to agricultural used grasslands, while the amount of
98 semi-natural habitats such as forests is generally spare (Mudri-Stojnić et al, 2012). Methodology of
99 sampling and site selection can be read in the original paper. For this study the sampled pollinators
100 were separated and aggregated per study-site into bees and non-bees such as hoverflies. The
101 information of forest cover in the vicinity was taken from available Pan-European forest cover maps
102 at a spatial resolution of 25m (Kempeneers et al., 2011). To get spatial information on the amount of
103 grassland and other land-cover types the CORINE land-cover dataset was obtained, clipped to the
104 respective region and resampled to a 25m resolution (EEA, 2013).

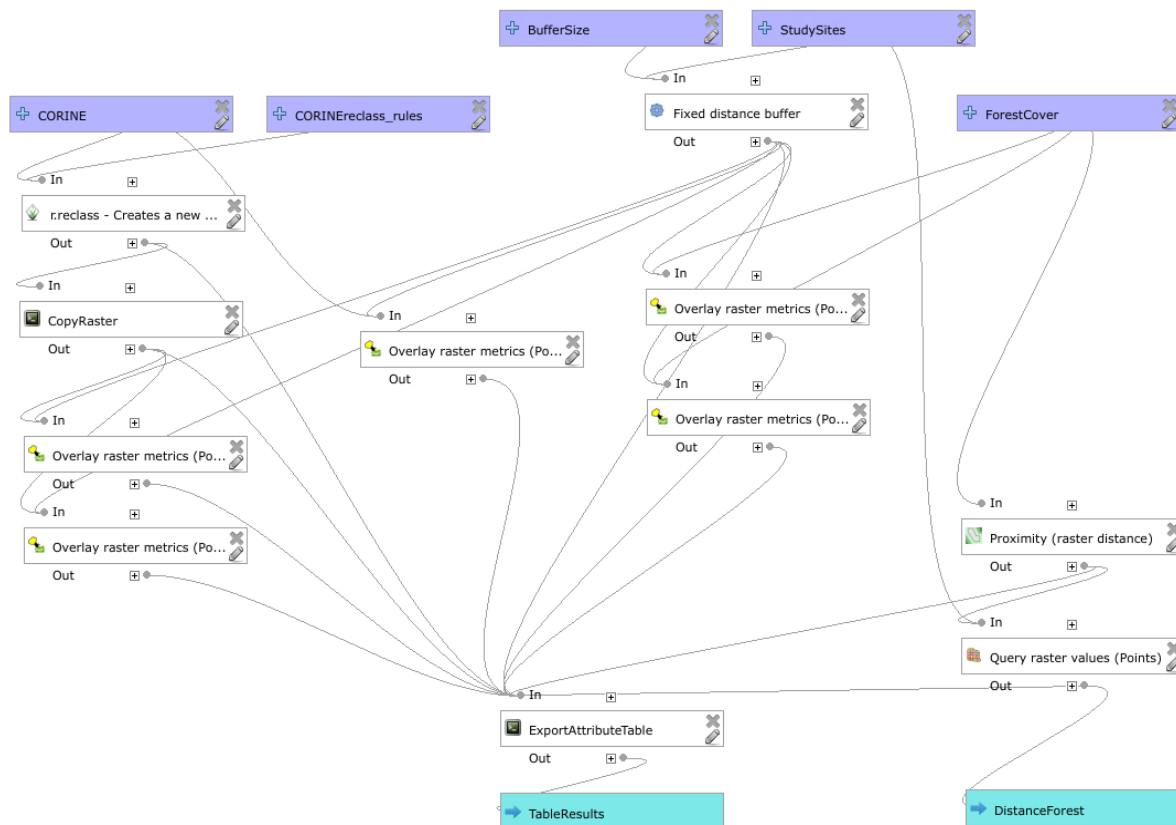


Figure 2: Shows the full model in QGIS to calculate a number of different landscape metrics for a given buffer size around the study sites. The full model was then executed in a batch-process with increasing buffer-size to calculate all metrics at various scales.

106 As potentially interesting predictors the percentage of forest cover (in %), edge density
 107 (edge length divided by number of forest patches), the total amount of agricultural used land in the
 108 vicinity (in m²), the number of wetland patches and the landscape heterogeneity were calculated. All
 109 predictors were estimated in a circular buffer around the individual study sites at various scales
 110 which were stepwise increased from 100m up to 3000m (100m per step) and thus represent a
 111 potential dispersal range (Jauker et al., 2009). Additionally the distance from the forest edge was
 112 calculated for each study site. All landscape analysis was done in a single model using LecoS and
 113 the QGIS processing model builder (Figure 2, Supplementary Data). The resulting data is then
 114 analyzed in R (R Core Team, 2013), by calculating a generalized linear model of the distance to the
 115 next forest against pollinator abundance and diversity. Furthermore the correlation coefficient of
 116 pollinator abundance and diversity against the landscape derived predictors was calculated for all
 117 buffer increments similar to other existing studies (Meyer et al., 2009).
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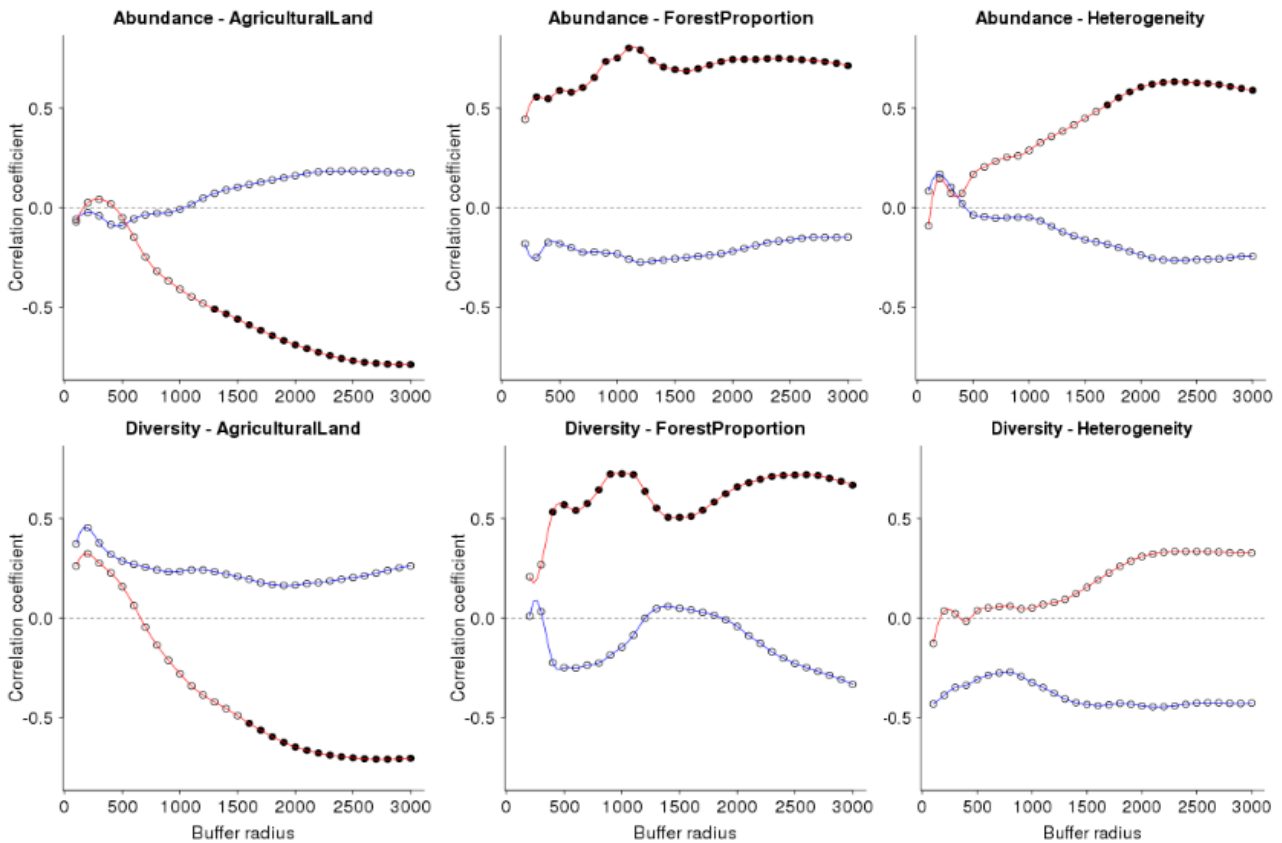


Figure 3: Pearson correlation coefficients between pollinator abundance and diversity and selected landscape metrics. Blue smoothed splines indicate bee species, while red splines show syrphid species. Coefficients are shown for all scales between 100m and 3000m buffer size. Filled circles denote significant correlations ($p < 0.05$) between the response and the predictor variable. Species richness and abundance was square-root transformed for analysis. The remaining figures can be found in the appendix.

120 Results and discussion

121 No significant relationship between the distance to main habitat at the study-site and bee or
 122 hoverfly diversity and abundance could be detected. However it could be shown that bee and
 123 hoverfly species indeed often have opposing responses at various landscape scales (Figure 3).
 124 Especially hoverfly abundance and diversity was significantly correlated with an increasing
 125 proportion of agricultural and forested land. Noteworthy is also the significant response of hoverfly
 126 abundance to increasing landscape heterogeneity at larger buffer scales (Figure 3). No significant
 127 correlation between any of the landscape metrics and bee abundance or diversity could be detected.

128 Although the pollinator survey conducted by Mudri-Stojnić et al, (2012) might be
 129 under-sampled and thus most likely does not fully reflect the whole existing species community, it
 130 is still able to show the often reported opposing difference between bees and hoverflies in response
 131 to landscape influences (Meyer et al., 2009; Jauker et al., 2009). Hoverfly abundance and diversity
 132 is supposed to be much more affected by land-cover change in the vicinity due to its dependency on
 133 microhabitats especially in forests (Branquart & Hemptinne, 2000). That might be the reason why

135 hoverfly diversity and abundance correlated so strongly with forest cover on the landscape level.

136 However the reader shall be reminded that this case-study can not be considered robust
137 evidence of general pollinator response to landscape influences due to its small sample size and
138 design simplicity. It is rather thought of as a demonstration of a possible study scenario where the
139 new QGIS plugin LecoS could potentially be used. In the authors opinion future studies should
140 especially focus on decomposing community responses to individual species responses to
141 landscape-wide resource provisioning. This has for instance already been done for some bumblebee
142 species, who surprisingly experience landscapes differently at various spatial scales (Westphal et al.,
143 2006).

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145 **Overview**

146 Here a new plugin for QGIS is presented which allows the computation of landscape-wide
147 and single patch metrics for use in ecological studies. The plugin itself is free and open-source and
148 can be modified and redistributed by potential users. Due to its functional integration into the
149 existing QGIS data processing framework it can be used in complex spatial models. The author
150 hopes that this plugin might be useful for ecologists and other people working with open-source
151 GIS products.

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