

Corrigendum to "Relationships between human activity and biodiversity in Europe at the national scale: Spatial density of human activity as a core driver of biodiversity erosion" [Ecol. Indicators 90 (2018) 356-365]

Frédéric Gosselin, Jean-Marc Callois

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Corrigendum to: Relationships between human activity and biodiversity in Europe at the national scale: spatial density of human activity as a core driver of biodiversity erosion (doi: 10.1016/j.ecolind.2018.03.010) Frédéric Gosselin¹ and Jean-Marc Callois² ¹Irstea, UR EFNO, Domaine des Barres, F-45290 Nogent-Sur-Vernisson, France. Mail: frederic.gosselin@irstea.fr ²Irstea, Département Territoires, 1 rue Pierre-Gilles de Gennes, F-92761 Antony Cedex, France. Mail: jean-marc.callois@agriculture.gouv.fr This corrigendum aims to correct the first three columns of Table 3 in Gosselin and Callois (2018), which was based on an erroneous program to calculate marginal Leave-one-out Information Criteria (LOOIC). Table 1 in the present paper replaces this previous Table 3. We discuss why the new results are more in agreement with the other results of Gosselin and Callois (2018) and why the marginal version of the LOOIC appears better than the conditional version. **Statistical methods** Here, the statistical models are the same as the ones in Gosselin and Callois (2018). They are Bayesian models which incorporate country random effects when observations were repeated in each country (for the two State Indicators and the Response Indicator), and which have a beta-binomial probability distribution for the two State Indicators and a zero-inflated beta distribution for the Response Inidcator and the two Pressure indicators. The zero-inflated beta distribution is a direct subproduct of the zero-inflated cumulative beta distribution that Herpigny and Gosselin (2015; MTUnlimited2 version) proposed to analyze plant cover class data. Indeed, we used statistical methods with a priori relevant probabilistic properties relative to our data. Based on the lessons

learned on partly similar data from Gosselin (2015), we paid careful attention to the potential over-dispersion of data as well as to the inclusion of random country effects.

We compared the models with the Leave-one-out Information Criterion (LOOIC) developed by

Vehtari et al. (2016). For State and Response indicators, we used the marginal version of the LOOIC

– i.e. the LOOIC integrated over the random effects –, for reasons discussed by Millar (2009, 2017),
and because the country random effect introduced a number of parameters which was of the same

order of magnitude as the number of observations. We calculated the marginal versions of the LOOIC

for 10,000 parameter values randomly drawn from the MCMC output by taking the mean of
probabilities of the observed data grouped by country over 1,000 random draws of the country

random effect. The original code in Gosselin and Callois (2018) was erroneous in that it did not take

into account the dependence between observations in the same country (i.e. the mean probability of
each observation was taken instead of the mean probability after grouping the observations of each
country). The new calculations used here correct this previous error.

41 Results

Model name	Extinct	Threatened	Protected	SEAL	iSEAL
	species	species	Areas		
Null	15.91	13.10	<u>0.00</u>	46.86	19.93
eco	4.37	11.66	7.34	18.26	0.00
leco	2.88	1.72	6.47	<u>0.00</u>	0.61

Kuznets	15.56	12.00	6.02	42.09	17.91
lKuznets	17.35	13.40	5.27	48.01	19.70
Kuznetsa	0.00	0.00	10.27	12.95	4.80
lKuznetsa	0.42	1.89	3.67	6.58	5.03
demo	5.16	10.51	11.94	17.16	0.82
ldemo	3.07	4.43	2.54	1.41	3.57
FC	1.91	4.05	4.45	40.82	22.14
SC	12.64	9.83	6.16	52.64	24.38
NGOs	14.72	15.28	5.24	39.10	15.05
ERV	11.00	14.11	3.12	44.85	22.32
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Table 1. Statistical comparison of models. Difference in Leave-one-out Information Criterion (LOOIC) values with the LOOIC of the best model fitted for the different explanatory models (by column). The lower the LOOIC, the better the model. The best model is underlined and models with a LOOIC relatively close to the best model—i.e. within six units of the best model—are in bold (as suggested in Millar, 2009). Only the models pertaining to the same target variable (models within the same column) are comparable. See Tables 1 and 2 in Gosselin and Callois (2018) for the content of the models. For the three first columns, the marginal version of the LOOIC is reported (see text).

51 Discussion

The new corrected values of the marginal LOOIC in Table 3 do not change any of the practical
conclusions in Gosselin and Callois (2018). However, they are more coherent with the levels of
significance of the estimators (Tables 4 to 7 in Gosselin and Callois 2018) than were the previous
values obtained (Table 3 in Gosselin and Callois 2018). Indeed, in the previous table, the levels of
significance of the most significant effects of explanatory variables were overall less significant for
the Proportion of Extinct and Threatened species (Tables 4 and 5 in Gosselin and Callois 2018) than
for the proportion of Sealed area or the dynamics of this proportion (Tables 6 and 7 in Gosselin and
Callois 2018). Yet, previous differences in LOOIC values with the null model had indicated stronger
differences for Extinct and Threatened species with respect to the dynamics in the proportion of
Sealed areas (Table 3 in Gosselin and Callois 2018).
The new model comparison results presented herein in Table 1 establish an ordering that is more in
line with the significance of the effects. This behavior is in agreement with McQuarrie and Tsai's
(1999) results; for orthogonal regressions, they showed a direct link between the statistical
significance of the effects of models and their Akaike Information Criterion (AIC), a criterion close to
LOOIC for such simple settings. Our settings are not very different from orthogonal regressions since
we mostly chose models with not too much correlated explanatory variables. The fact that the new
marginal LOOIC are more coherent with the levels of significance of the estimators than were the
previous values obtained is therefore welcome. In the case of the SC model for the proportion of
Threatened species (Table 1), the results for marginal LOOIC shown in Table 1 are also more
satisfactory than the conditional LOOIC results presented in Table SM1 (Supplementary material):
given the two significant estimators involved in the model (Table 5 in Gosselin and Callois 2018),
marginal LOOIC indicates that the SC model is better than the Null model, qualitatively in agreement
with McQuarrie and Tsai (1999)'s results, while conditional LOOIC indicates that it is not as good as
the Null model.

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