

INSTITUTO NACIONAL DE PESQUISAS DA AMAZÔNIA - INPA
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**Limitations in the use of species distribution models for environmental
impact assessments in the Amazon**

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Manaus, Amazonas
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**Limitations in the use of species distribution models for environmental impact
assessments in the Amazon**

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Sinopse

Intensamente utilizada e debatida pela comunidade científica nos últimos anos, a modelagem de distribuição de espécies foi apontada como uma potencial ferramenta para direcionar decisões político-administrativas no contexto do licenciamento ambiental de empreendimentos no Brasil e na Amazônia. Nesse estudo utilizamos dados reais coletados na avaliação de impacto ambiental de um grande aproveitamento hidrelétrico localizado no Rio Madeira – Rondônia – Brasil, para avaliar as limitações e potencialidades do uso das técnicas de modelagem no direcionamento de decisões no contexto do licenciamento na região. Os resultados sugerem que, para viabilizar abordagens utilizando modelos de distribuição de espécies alvo no auxílio de demandas voltadas à avaliação de impacto no bioma Amazônico é necessário, ou expandir as áreas de amostragem dos estudos obrigatórios prévios, definida pelo órgão ambiental regulador, ou investir em estratégias de levantamentos integrados e comparativos da biodiversidade na região, com o objetivo de melhorar o cenário amostral e suprir os pré-requisitos do uso das técnicas.

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Aos pesquisadores Bárbara Zimbres, Renata Dias Françoso, Yuri Salmona (Laboratório de Planejamento para Conservação da Biodiversidade (UnB), Victor J. T. Perdigão (Eletronorte), Stéphanie Watzel, André Luis Sousa, Darren Norris, Pedro Ivo Simões, Igor Kaffer (UFAM), Sérgio Henrique (MMA) e David Cho (IBAMA) pelo apoio na construção das idéias e das análises. À David Tonks, Luciene Robichaud, Stéphanie DeAlmeida pelas revisões de tradução.

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Limitações no uso da modelagem de distribuição de espécies no contexto das avaliações de impacto ambiental de empreendimentos na Amazônia.

Resumo: A modelagem preditiva de distribuição de espécies foi apontada como uma potencial ferramenta para direcionar decisões nos processos de licenciamento ambiental de empreendimentos na Amazônia. Nesse manuscrito apresentamos um estudo de caso utilizando dados da anurofauna levantada no estudo de impacto ambiental de impacto de um aproveitamento hidrelétrico localizado no Rio Madeira – Rondônia – Brasil, para avaliar se o uso dos MDEs é adequado para orientar decisões nos processos de licenciamento na região. Como as estratégias de conservação envolvem a priorização de alvos, primeiramente definimos as espécies da anurofauna (Ordem Anura) prioritárias para as ações de manejo e analisamos suas respectivas situações amostrais. Posteriormente, à luz do que é previsto na legislação ambiental do país e diante as recomendações descritas na literatura especializada sobre o uso das ferramentas, discutimos as limitações e potencialidades da utilização dessas técnicas para orientação de ações mitigatórias e compensatórias na região. Os resultados mostraram que a maior parte das espécies alvo não dispuseram de dados de presença representativos para o treinamento dos modelos, pois elas são, em sua maior parte, ainda não descritas. Logo, o uso dos algoritmos somente seria indicado utilizando uma parcela dos alvos selecionados ou com táxons não prioritários. Considerando as inúmeras lacunas de conhecimento sobre a distribuição dos táxons na Amazônia, e ainda que os inventários de fauna no contexto do licenciamento são tipicamente realizados nas adjacências da área prevista para ser impactada, os modelos de distribuição de espécies precisariam extrapolar os registros de presença dos táxons alvo para assim orientar nas decisões de manejo em ampla escala. Os resultados das simulações que extrapolaram os dados geograficamente corroboraram diversos autores que sugerem a diminuição da robustez das previsões com o aumento da extensão das extrações. Nós concluímos que o uso dos modelos nesse contexto requer a expansão das áreas amostradas nos estudos de impacto ou que haja investimentos em estudos estratégicos que forneçam dados integrados e comparativos no âmbito do bioma para melhorar a situação amostral na região e suprir os pré-requisitos do bom uso dessas técnicas.

Limitations in the use of species distribution models for environmental impact assessments in the Amazon

Abstract: Species distribution models (SDMs) have been recognized as a potential tool for guiding decisions regarding the process of environmental licensing in the Amazon. In this paper, we present a case study using data on frogs that were obtained from the impact assessment of a hydroelectric project located on the Rio Madeira, Rondônia, Brazil, to evaluate the use of SDMs in the decision-making process for licensing works in the Amazon Basin. Because conservation strategies must prioritize targets, we defined the priority species for management actions and analyzed their respective sampling situations. Based on the expectations of environmental legislation in Brazil and using the tools recommended in the literature, we discussed the limitations and potential use of SDM techniques for guiding mitigation actions in the region. The results showed that there were insufficient data for most of the target species to be included in the distribution models because most species have not yet been described. Because there are numerous gaps in the Brazilian samples and the current mandatory surveys are typically conducted in areas adjacent to the affected areas, the models must extrapolate beyond the sampled data to guide decisions, such as defining areas to be used to offset the effects. The results of geographical extrapolation simulations, using species collected in the affected area, were corroborated by several authors who suggested that the robustness of predictions decreases with the extension of the extrapolations. We conclude that the use of SDMs for supporting decisions to license projects in the Amazon requires expanded sampling areas for impact studies or an investment in integrated and comparative survey strategies to improve biodiversity sampling and to fulfill the prerequisites for the use of such techniques.

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Introdução geral

O amparo científico é reconhecidamente importante no subsídio de decisões político-administrativas no âmbito das questões ambientais de um país. No Brasil, um exemplo dessa sinergia entre pesquisa e deliberação governamental previsto por lei, se refere aos processos de licenciamento ambiental. Nesses processos, os órgãos ambientais públicos buscam conciliar a instalação e funcionamento de atividades potencialmente causadoras de impacto com a conservação do meio ambiente. Para isso, estudos multidisciplinares prévios são solicitados aos interessados que desejam instalar empreendimentos ou renovar suas licenças de operação.

Sendo os estudos ambientais desenhados, orientados e analisados pelos órgãos ambientais governamentais, cabe à comunidade científica a incumbência do acompanhamento e discutir as abordagens e técnicas utilizadas pelos agentes públicos, nas mais diversas deliberações referentes ao manejo ou conservação da biodiversidade. Nesse sentido, esse manuscrito teve como objetivo avaliar a viabilidade do uso das ferramentas de modelagem de distribuição de espécies, amplamente utilizada e discutida pela comunidade científica nos últimos anos, no subsídio de decisões no contexto da avaliação de impacto de empreendimentos na Amazônia. Com os resultados esperamos contribuir com o aprimoramento dos processos de comando e controle de impacto, exercidos pela poder público brasileiro.

Objetivos

Objetivo geral:

Avaliar o desempenho de modelos de distribuição de espécies no orientação de questões práticas no contexto do licenciamento ambiental de empreendimentos na Amazônia.

Objetivos específicos:

1. Avaliar a situação amostral dos táxons alvo na parametrização dos modelos;
2. Avaliar o desempenho dos modelos em ocasiões de extração e interpolação geográfica e ambiental dos dados de presença dos alvos.
3. Orientar tomadores de decisões sobre as limitações e potencialidades do uso das ferramentas de modelagem de distribuição no auxílio de diferentes demandas do licenciamento;

Capítulo 1

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1 **Limitations in the use of species distribution models for environmental impact
2 assessments in the Amazon**

3

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17

18 **Abstract:** Species distribution models (SDMs) have been recognized as a potential tool
19 for guiding decisions regarding the process of environmental licensing in the Amazon.
20 In this paper, we present a case study using data on frogs that were obtained from the
21 impact assessment of a hydroelectric project located on the Rio Madeira, Rondônia,
22 Brazil, to evaluate the use of SDMs in the decision-making process for licensing works
23 in the Amazon Basin. Because conservation strategies must prioritize targets, we
24 defined the priority species for management actions and analyzed their respective
25 sampling situations. Based on the expectations of environmental legislation in Brazil
26 and using the tools recommended in the literature, we discussed the limitations and
27 potential use of SDM techniques for guiding mitigation actions in the region. The
28 results showed that there were insufficient data for most of the target species to be
29 included in the distribution models because most species have not yet been described.
30 Because there are numerous gaps in the Brazilian samples and the current mandatory
31 surveys are typically conducted in areas adjacent to the affected areas, the models must
32 extrapolate beyond the sampled data to guide decisions, such as defining areas to be
33 used to offset the effects. The results of geographical extrapolation simulations, using
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36 extrapolations. We conclude that the use of SDMs for supporting decisions to license
37 projects in the Amazon requires expanded sampling areas for impact studies or an
38 investment in integrated and comparative survey strategies to improve biodiversity
39 sampling and to fulfill the prerequisites for the use of such techniques.

40

41 **Keyword:** EIA/RIMA, environmental licensing, fauna surveys, Hydroelectric
42 developments, species distribution

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45 **Limitações no uso da modelagem de distribuição de espécies no contexto das
46 avaliações de impacto ambiental de empreendimentos na Amazônia.**

47

48 **Resumo:** A modelagem preditiva de distribuição de espécies foi apontada como uma
49 potencial ferramenta para direcionar decisões nos processos de licenciamento ambiental
50 de empreendimentos na Amazônia. Nesse manuscrito apresentamos um estudo de caso
51 utilizando dados da anurofauna levantada no estudo de impacto ambiental de impacto
52 de um aproveitamento hidrelétrico localizado no Rio Madeira – Rondônia – Brasil, para
53 avaliar se o uso dos MDEs é adequado para orientar decisões nos processos de
54 licenciamento na região. Como as estratégias de conservação envolvem a priorização de
55 alvos, primeiramente definimos as espécies da anurofauna (Ordem Anura) prioritárias
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57 Posteriormente, à luz do que é previsto na legislação ambiental do país e diante as
58 recomendações descritas na literatura especializada sobre o uso das ferramentas,
59 discutimos as limitações e potencialidades da utilização dessas técnicas para orientação
60 de ações mitigatórias e compensatórias na região. Os resultados mostraram que a maior
61 parte das espécies alvo não dispuseram de dados de presença representativos para o
62 treinamento dos modelos, pois elas são, em sua maior parte, ainda não descritas. Logo,
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65 conhecimento sobre a distribuição dos táxons na Amazônia, e ainda que os inventários
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67 área prevista para ser impactada, os modelos de distribuição de espécies precisariam
68 extrapolar os registros de presença dos táxons alvo para assim orientar nas decisões de
69 manejo em ampla escala. Os resultados das simulações que extrapolaram os dados
70 geograficamente corroboraram diversos autores que sugerem a diminuição da robustez
71 das previsões com o aumento da extensão das extrações. Nós concluímos que o uso
72 dos modelos nesse contexto requer a expansão das áreas amostradas nos estudos de
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74 integrados e comparativos no âmbito do bioma para melhorar a situação amostral na
75 região e suprir os pré-requisitos do bom uso dessas técnicas.

76

77

78 **Palavras-chave:** EIA/RIMA, licenciamento ambiental, levantamentos de fauna,
79 empreendimentos hidrelétricos, distribuição de espécies

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87 **1. Introduction**

88

89 Species distribution models (SDMs) have many potential uses for conservation
90 planning (Araújo et al. 2004, Ferrier 2002, Lopez-Lopez et al. 2007, Traill and Bigalke
91 2006). Tôrres and Vercillo (2012) suggested using these tools to guide decisions related
92 to environmental licensing in Brazil, which is considered an important instrument for
93 the country's environmental policy and is a mandatory government control procedure
94 that is enforced prior to the installation of facilities or activities that may interfere with
95 the environment (Law No. 6.938/1981, article 10, of Law Decree No. 99.274/1990 art.
96 17, Res CONAMA No. 237/1997, Res CONAMA No. 001/1986). There are three
97 distinct stages that occur prior to granting an environmental permit: preliminary
98 licensing, installation licensing and operational licensing. In each stage, the contractor
99 must conduct specific studies, while the government agencies assess the environmental
100 feasibility of the project and make decisions concerning actions required to mitigate
101 environmental damage (Complementary law No 140/2011).

102 In addition to mitigation measures that seek to minimize the negative effects, the
103 licensing process also provides environmental compensation for the permanent or
104 temporary use of natural resources that become unavailable (Giasson and Carvalho,
105 2012). These environmental offsets can be used for land regularization or for the
106 implements or maintains protected areas, which are selected according to priorities and
107 guidelines set by the Federal Board of Environmental Compensation - CFCA (MMA
108 Ordinance No. 416/2010). Based on these guidelines, the Federal Environmental
109 Compensation Committee (CCAF) defines the allocation of compensation (Decree Law
110 No. 4.340/2002, Decree No. 6.848/2009).

111 Tôrres and Vercillo (2012) suggested that SDMs should be used by licensing
112 agencies as potential tools to improve environmental study requirements, reduce the
113 subjectivity of impact analyses, and improve mitigation and compensation measures.
114 Therefore, SDMs could indicate areas that are environmentally appropriate or
115 inappropriate for the target species and assist in management actions, such as locating
116 sample sites, relocating sites that are expected to be impacted, and selecting reserves to
117 offset the damage. Several authors have indicated limitations in the use of SDMs,
118 especially when the amount of data is insufficient for parameterizing the models
119 (Kadmon et al. 2003, Stockwell and Peterson 2002, Wisz et al. 2008), whereas others
120 have questioned the robustness of the predictions based on extrapolations (Elith and
121 Leathwick 2009, Miller et al. 2004). However, no studies have analyzed the use of
122 SDMs to support decisions regarding environmental licensing in tropical regions.

123 In this paper, we evaluate a case study using data collected during the
124 environmental licensing of a hydroelectric plant constructed in the Amazon Basin to
125 assess whether SDMs are appropriate for guiding management requirements in the
126 region based on the type and quality of information that is typically generated by
127 environmental impact studies. We do not discuss the conceptual or theoretical
128 evaluations of species modeling tools' functions or the effectiveness of theoretical
129 modeling techniques because these topics have been extensively discussed in the
130 literature (Araújo and Guisan 2006, Elith and Graham 2009, Elith et al. 2011, Royle et
131 al. 2012). Instead, our goal was to evaluate whether the species modeling tools are
132 adequate to guide decisions related to environmental licensing in the Amazon based on
133 the requirements described in the literature for the proper functioning of the modeling
134 techniques.

135

136 **1.1 Case study**

137

138 The Santo Antônio hydroelectric plant on the upper Madeira River is located
139 approximately seven kilometers from the city of Porto Velho, Rondônia State, and
140 began operation in March 2012. As part of the licensing process, an investigation of the
141 local herpetofauna was conducted between February 2010 and November 2011; five
142 field surveys were undertaken. The objective of the surveys was to identify the
143 herpetofauna in the area that may be impacted by the project and to quantify the
144 potential effects of directly filling the dam on the sampled species and biological
145 communities. The sample design was defined by a team of consultants, scientists
146 responsible for the study, and public officials from the Brazilian Institute of
147 Environment and Renewable Natural Resources according to legal guidelines, which
148 require an environmental assessment of the affected area (art. 6 of CONAMA
149 Resolution No 001/86).

150 Because of the time and financial constraints involved in implementing
151 mitigation actions, conservation strategies should target the most threatened species
152 (Loyola and Lewinsohn 2009). Therefore, the first task was to identify target species for
153 management actions and to verify the availability of occurrence data for those species to
154 calibrate the model at the scales at which decisions need to be taken.

155 The licensing agency may require local- to national-scale assessments depending
156 on the decision scenario and extent of the effects (minutes from regular meeting No. 9
157 of CCAF-10/2012, Complementary law No 140/2011, CONAMA Resolution No.
158 237/1997). For example, models at a local scale might be useful for the installation of

159 sample sites to determine the distribution of species threatened by the project.
160 Moreover, models based on large catchments (federal environmental planning units)
161 might be used to investigate the biological irreplaceability of environments expected to
162 be affected. Thus, we considered three spatial scales in our analyses. The direct
163 influence area of the project (AID) is considered to be the local spatial scale; the
164 Amazon hydrographic region, which is used as a planning and management unit (Law
165 No. 9.433/97) to meet the requirements of impact relocation decisions, is considered the
166 second spatial scale; and the mosaic of protected areas that could receive indemnities as
167 a result of the effects of the project is considered the third spatial scale.

168 In this study, we only considered amphibious species as target species because
169 they are frequently used as bioindicators of human disturbances in terrestrial and
170 aquatic ecosystems (Barros 2005, Heyer et al. 1994). Moreover, two *Allobates* species
171 in the project area could be used to evaluate the model results because data were
172 available on their presence and absence in areas outside the area affected by the dam.

173

174 **1.2 Species distribution models for impact assessment decisions**

175

176 Most species distribution algorithms attempt to identify locations with similar
177 environmental conditions to those at sites where the taxa of interest are known to occur
178 (Saupe et al. 2012). Royle et al. (2012) suggested that models that estimate the
179 probability of occurrence would be more robust and ecologically coherent. Although
180 their technique is compatible with presence-only records, the authors emphasized the
181 need for random sampling within species distributions for estimates produced by formal
182 quantitative methods (Royle et al. 2012).

183 When data are available throughout the range of the species, occurrence
184 information can be interpolated. Moreover, in principle, it should be possible to estimate
185 both the probability of occurrence (Royle et al. 2012) and environmental suitability
186 (Elith et al. 2011). However, in tropical regions, information on the distribution of most
187 taxa is rare, is usually concentrated in specifically selected locations, and is
188 systematically sampled (Cayuela et al. 2009, Feeley and Silman 2011, Moerman and
189 Estabrook 2006).

190 Elith et al. (2011) also suggested that using random sampling within the
191 distribution of the species is preferable in approaches that use index-based methods
192 (Elith et al. 2006, Elith et al. 2011, Hernandez et al. 2006, Wiza et al. 2008). However,
193 to identify areas where target species may occur, algorithms may be required to
194 extrapolate predictions beyond the sampled areas in impact studies. Geographical
195 extrapolations beyond the known area of occurrence may or may not involve
196 extrapolation beyond the range of environmental variables used in the construction of
197 the models. In this situation, estimating the probability of occurrence is not possible, but
198 the environmental suitability is possible (Elith et al. 2011).

199 Models must extrapolate or interpolate information gathered in the area of
200 biological surveys (AID) so that their predictions support management decisions at
201 different spatial scales. In this study, we assessed whether the products of
202 environmental suitability models are appropriate to guide the type of management
203 concerns identified by Tôrres and Vercillo (2012), which require geographically
204 interpolated or extrapolated data of the target species. Knowledge of the practical
205 limitations of distribution models will assist environmental agencies and consultants

206 when planning management decisions and future sampling design studies in tropical
207 regions.

208

209 **2. Material and methods**

210

211 **2.1 Target species selection**

212

213 Technical evaluations of the fauna in Brazilian environmental assessments are
214 based on biological surveys undertaken in the project's affected area (IBAMA
215 Normative Instruction No. 146/07, CONAMA Resolution No. 001/86). The fauna
216 survey undertaken for the Santo Antônio licensing was conducted in the area of direct
217 impact (AID). Therefore, we consider these limits as the local scale for determining the
218 requirements of environmental studies (Fig 1). A central objective of these surveys was
219 to support the definition of management and conservation strategies for species
220 threatened by the installation and operation of the project (target species).

221 Current Brazilian law does not specify criteria for selecting targets in
222 environmental impact studies. However, the CONAMA Resolution No. 001/86, article
223 6°, and the IBAMA Normative Instruction No. 146/07 (restricted to the licensing of
224 hydroelectric power plants by the IBAMA Normative Instruction No. 10/2009, article
225 5°) stipulate that species that indicate environmental quality and scientific or economic
226 value and are endangered or rare must be highlighted in the results of environmental
227 evaluations for mitigation efforts.

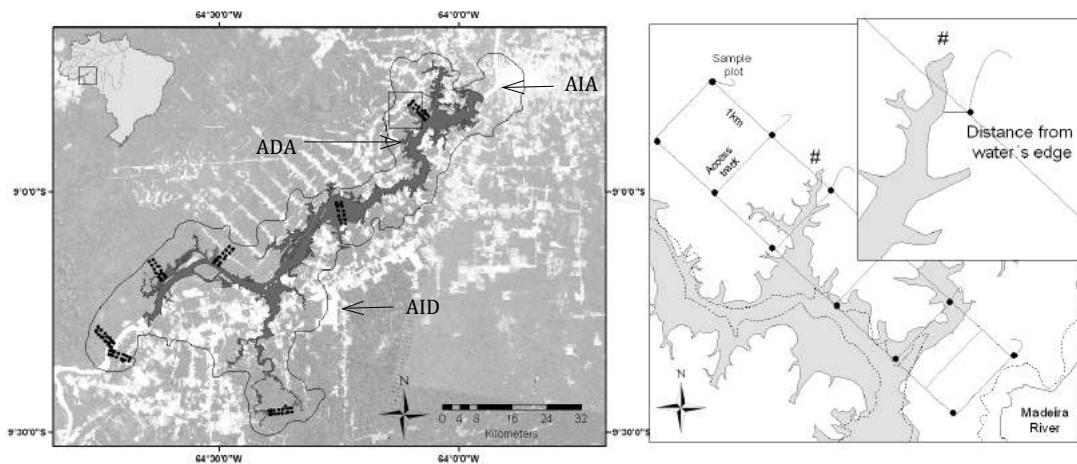
228 However, the lack of knowledge regarding the biodiversity in the Amazon limits
229 the classification of most organisms detected in surveys in the categories required by

230 law. Numerous Amazonian species remain poorly understood. Moreover, many species
231 are categorized as “data deficient” by the International Union for Conservation of
232 Nature (IUCN). Others species that are considered threatened on a national scale are
233 often abundant in the Amazon biome. Therefore, we used data gathered during the
234 surveys for our target selection and focused on species with limited known distributions
235 outside the AID.

236 The sampling design used in the environmental impact studies was based on the
237 RAPELD method (Costa and Magnusson 2010, Magnusson et al. 2005, Magnusson et
238 al. 2013), which was adapted to measure the impact of dam flooding on the spatial
239 distribution of terrestrial organisms. Modules with two trails perpendicular to the
240 original course of the river were systematically distributed in the area expected to be
241 affected by the project (ADA). The arrangement of the modules perpendicular to the
242 river was used to assess the impact of the flooding gradient on the detected species. In
243 each trail, 250-m-long permanent plots were installed at 1 km intervals beginning on the
244 riverbank. Extra plots were installed at 0 and 1000 m to better evaluate the riparian
245 areas. The plots followed the altitudinal contours to minimize the internal variation in
246 the distance to the water table and soil; therefore, these variables can be used to predict
247 species distributions (Figure 1). Frogs were sampled using an active search technique
248 that was limited by time, pitfall traps installed in permanent plots, and random
249 encounters along the access trails (details of the methods are available in Lima et al.,
250 2011; Technical Information No. 65/2008 - COHID / CGENE / DILIC / IBAMA).

251 The selected target species were exclusively detected in the area that was
252 expected to be flooded by the reservoir (ADA). We used the smallest distance between
253 the estimated maximum level of flooding and local detection of each specimen to

254 measure the degree of threat. Values equal to 0 indicated points located at the edge of
 255 the expected flooding zone, negative values refer to the distance (meters) between
 256 species observations and the flooding edge in the direction toward the center of the dam,
 257 and positive values indicate distances from the edge of the dam away from flooded
 258 areas.



259
 260
 261 **Figure 1.** Left side: a map of the distribution of sample modules (groups of points) in the area
 262 of direct impact (the thin line represents the buffer around the impacted area). The maximum
 263 flood elevation resulting from the filling of the reservoir represents the directly affected area
 264 (dark gray area). Right side: a schematic drawing of a sample module positioned perpendicular
 265 to the Madeira River in which the points represent the beginning of the permanent plots. The
 266 inset shows an example of a distance considered in the analysis for selecting the target species.
 267 Species that were only detected in the submerged area were considered targets because they are
 268 directly threatened by the filling of the dam.
 269
 270

271 Management strategies depend on the spatial scale of the analysis (Hortal et al.
 272 2010). Therefore, species widely distributed in the Amazon might be impacted locally
 273 by the installation of an enterprise; however, these species might not be threatened over
 274 larger areas. Other species might be threatened at all spatial scales if they are only
 275 detected in impacted areas. To filter the criteria for prioritizing management actions
 276 based on the biological complementarity (Caro 2010) between the area expected to be
 277 flooded (directly affected area - ADA) and areas adjacent to the impacted area

278 (indirectly affected area - AIA), we also considered the complementarity of species
279 distributions between the ADA and other surveyed regions of the Amazon biome. To
280 evaluate the complementarity of the ADA with areas distant from those directly
281 sampled, we sought occurrence records of target taxa defined at the local scale in
282 collections provided by museums and online databases (SpeciesLink, GBIF, IUCN,
283 Encyclopedia of Life) and by consulting experts.

284

285 **2.2 Species distribution models for impact assessment decisions**

286

287 Species distribution models were interpolated and extrapolated in both
288 geographic and environmental space. Geographic space refers to the broad geographic
289 area between the target detection area and the area of inference as defined by
290 management concerns. Such inferences in geographic space can also add environmental
291 extrapolations between distant or neighboring regions if there are large variations in
292 temperature, humidity, fauna, and flora. Extensive environmental extrapolation in which
293 the training data are not representative of the environmental conditions in the area of
294 inference increase model uncertainties (Elith et al. 2006, Pearson et al. 2006, Thuiller et
295 al. 2004).

296 To evaluate model performance when conducting geographical or environmental
297 extrapolations and interpolations, we divided the analysis into two parts. The first part
298 evaluates the environmental extrapolation models (which were generated with the target
299 species) between the species occurrence area and the region of management concern. In
300 the second analysis, we used reliable presence and absence data for two species with
301 known geographic distributions to evaluate the performance of the models when

302 interpolating within or extrapolating outside of the geographical limits of the training
303 data.

304 The maximum entropy algorithm (MaxEnt) of the geographic distribution of
305 species (Phillips et al. 2006) generally performs better than alternative procedures when
306 sampling limitations are in effect (Elith et al. 2006, Elith and Leathwick 2009,
307 Hernandez et al. 2006, Wiza et al. 2008). Maximum entropy refers to a type of
308 statistical inference arising from information theory and provides a constructive
309 criterion for determining probability distributions based on partial knowledge. The
310 algorithm allows approaches with only presence data and predictions by interpolation or
311 extrapolation; therefore, we used the MaxEnt algorithm to parameterize the models used
312 in the analyses.

313 Because the environmental predictors change according to the spatial scale
314 (local to regional) of the analysis (Hortal et al. 2010), each model used different layers
315 that could reflect ecological requirements, had strong relationships to the data (selected
316 according to a jackknife test), and were not strongly correlated (Pearson correlation \leq
317 0.8) (Elith et al. 2006, Elith et al. 2010). Detailed variables at the local scale and
318 measures of the interactions between species were not available for inclusion in the
319 models.

320

321 **2.2.1 Effects of environmental interpolations and extrapolations on**
322 **management requirements**

323

324 Elith et al. (2010) developed a measurement of environmental similarity that
325 they called multivariate environmental similarity surfaces (MESS). MESS are used to

326 alert and minimize errors in extrapolation approaches with species distribution models
327 (see technical details in Elith et al. 2010 - Appendix S3). The MESS analysis is similar
328 to the BIOCLIM algorithm (Busby 1991); however, MESS analysis uses percentiles
329 that are extended to produce negative values referring to sites where at least one
330 variable has a value that is outside the environmental range of the reference set (Elith et
331 al. 2010).

332 We generated models with target species data to assess the extent of
333 extrapolation in two areas of interest and to assess whether the environments are similar
334 or different between the area of interest and environments where the target species have
335 been sampled regardless of their suitability (Saupe et al. 2012). One area of interest is
336 the directly impacted area (AID), wherein the models assist in determining the
337 distribution of targets; the other area of interest is the mosaic of protected areas selected
338 by the Federal Environmental Compensation Committee (minutes from regular meeting
339 No.10/2012). In total, 21 protected areas under municipal, state, and federal government
340 administration were compensated for maintenance purposes (minutes from regular
341 meeting No. 9 of CCAF-10/2012). To generate MESS maps, we prepared the reference
342 masks with an ASCII grid. The value “1” represented pixels used in training the models
343 with occurrence data. Moreover, “NoData” was used outside the model area (Elith et al,
344 2010).

345

346 **2.2.2 Effects of geographical interpolations and extrapolations**

347

348 We used data from two species of the family Aromobatidae that were detected in
349 the surveys of the area affected by the dam for which presence and absence data were

350 available within and outside the area affected by the project (Caldwell and Lima 2003,
351 Kaefer et al. 2012, Lima et al. 2007, Lima et al. 2010, Simões et al. 2008, Simões et al.
352 2010). Although the two species were not considered targets in this study, they allowed
353 us to simulate limited sampling and thus evaluate the model predictions in cases where
354 the algorithm must interpolate and extrapolate predictions to guide environmental
355 compensation. Furthermore, many taxa in this family are included as targets in
356 environmental impact studies because they have restricted distributions, do not migrate,
357 are territorial and short-lived, and have restricted gene flow between adjacent sites
358 (Kaefer et al. 2013). Therefore, environmental perturbations in their areas of occurrence
359 may be considered threats to the species.

360 The effect of incomplete sampling on the prediction of environmentally
361 appropriate areas by extrapolation and interpolation was measured by dividing the
362 distribution data for both species into three macro-scale groups (Amazon Basin) and
363 two local-scale groups (AID). Each group had a similar number of records, so that, at a
364 local scale, a partition was used for training and the other for test. To increase the
365 number of records for model construction at the macro-scale, we used two groups for
366 training each species and one for testing.

367 The algorithm performance was measured using the AUC test values (the area
368 under the curve of a graph characteristic of the receiver) as a measure of the failure rate.
369 The models were also run at each spatial scale with all available occurrences for each
370 species, with 20% of the occurrences randomly assigned by the algorithm to test the
371 models. The average of the AUC test results for 10 random arrangements was used as
372 result.

373 Because we considered predictions of environmental suitability, commission
374 errors were not considered inappropriate. However, such errors are not desirable in
375 publicly subsidized actions because if resources for further surveys are not available,
376 then agencies will be more likely to interpret the results as the probability of occurrence
377 rather than potential environmental suitability. To better visualize the maps of
378 environmentally appropriate areas, each image (raster map) prediction was imported
379 into ArcGIS 9.3 to reclassify the images based on the “minimum training presence”
380 thresholds; the differences between the predicted areas and the known species
381 distribution ranges were illustrated and discussed. Interpretations were based on a
382 simple heuristic scheme for the BAM diagram (Soberón and Peterson 2005, Soberón
383 2007), which summarizes the joint effects of external biotic and abiotic characteristics
384 and dispersal on species distribution predictions.

385 The species selected for the simulations were *Allobates nidicola* (Lima et al.
386 2007), which was recently described and has a distribution restricted to the Purus-
387 Madeira interfluve, and *Allobates femoralis* (Boulenger 1883), which is widely
388 distributed. Similar to many Amazonian species with a wide distribution, *A. femoralis*
389 may be a species complex; however, the classification follows the current state of
390 knowledge regarding the taxon (Amézquita et al. 2009, Simões et al. 2010). Occupancy
391 estimates and detection for these species were performed using data from sampling
392 repetitions in different monitoring campaigns through hierarchical models that
393 considered imperfect detection (Mackenzie et al. 2003) and using the Presence software
394 (Hines 2006). Because of the known demographic characteristics of species in this
395 family, we used the “single species” model with colonization and extinction parameters
396 held constant.

397 The probabilities of detection (ρ) for individual surveys for both taxa (Table 2)
 398 were sufficiently high; therefore, we are confident that the absence records for areas that
 399 have been repeatedly surveyed, such as those used in our modeling component,
 400 indicated the absence of the species from the area.

401

402 **Table 1:** Summary of the models and parameter estimates of *A. femoralis* and *A. nidicola*,
 403 including the values of the Akaike information criterion (AIC), number of parameters (k),
 404 occupancy (Ψ), and detection (ρ).
 405

Species	Model	AIC	-2log (likelihood)	K	Ψ	Std. err	95% conf. Interval	ρ	Std. err	95% conf. interval
<i>A. femoralis</i>	psi(.),gamma(.),eps(.),p(.)	321.382	317.382	2	0.57	0.06	0.45-0.68	0.59	0.04	0.51-0.67
<i>A. nidicola</i>	psi(.),gamma(.),eps(.),p(.)	161.639	157.639	2	0.27	0.07	0.16-0.43	0.31	0.07	0.19-0.47

406

407 **3. Results**

408

409 **3.1 Target species selection**

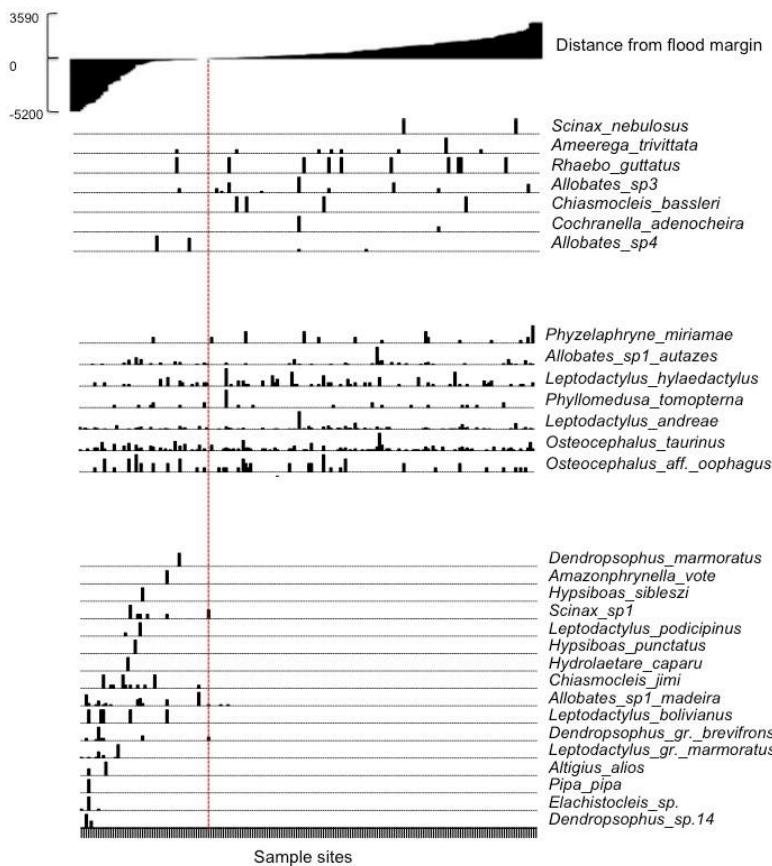
410

411 The anuran species found in the reservoir's affected area formed three groups
 412 that had different management priorities (Figure 2). The first group included species
 413 that were only found in locations distal to large bodies of water; therefore, they were far
 414 from the areas that were impacted by the project. The second group was composed of
 415 taxa detected close to and far away from the area to be flooded. A third group included
 416 species that were only found in areas where flooding was expected. Species in the third
 417 group were considered potentially threatened by the project and therefore were
 418 considered target species. Among the detected target species, eight species were
 419 previously recorded elsewhere in the Amazon Basin, far from the dam location. For

420 these species, the impact of the dam on the taxa was restricted to the scale of the project
421 site (Table 1). One species has only been observed outside of Brazil (*Altigius alios*).
422 Moreover, *Hidrolaetare caparu* has only been detected in one other location in Brazil
423 and is typical of Bolivia (Ferrão and Lima - personal communication). Accordingly,
424 these species are important targets for national biological heritage. Seven of the target
425 taxa were potentially undescribed species and would be considered target species
426 (scientific value) based on current legislation. Other potentially undescribed species
427 were also detected in areas distal to the main affected area (e.g., *Allobates* sp3) or
428 sampled widely across the analyzed spatial gradient (*Allobates* sp1 *autazes*,
429 *Osteocephalus* aff. *oophagus*). The macro-scale distributions of these species are not
430 available.

431 None of the detected target species were classified as threatened by the IUCN
432 Red List or the official list of Brazilian threatened fauna species (MMA Instruction No.
433 03/2003). Moreover, none of the target species had economic value.

434



435

436 **Figure 2:** Relative number of individuals detected per plot in relation to the distance of the plot
 437 (m) from the edge of the expected flood area (impact of the dam flooding - predicted maximum
 438 quota = 70.5 m). Data on which the figure is based are provided in Appendix 1.

439

440

441 **Table 2:** Diagnosis of species occurrence and complementarity between the area directly
 442 affected by the dam and areas in the state of Rondônia (RO), other areas in the Legal Brazilian
 443 Amazon, and the entire Amazon, which includes areas in other countries. “X” represents
 444 regions where the species have been previously recorded. Question marks indicate regions in
 445 which the species have not been detected. *According to Ferrão and Lima (personal
 446 communication), *Hydrolaetare caparu* was also recorded by M. Gordo in the Corumbiaria State
 447 Park; its typical locality is Bolivia.

448

449

Target species	Independent occurrence records	Status IUCN	State (RO)	Brazilian Amazon	Foreign Amazon
<i>Dendropsophus marmoratus</i>	1	Least Concern	X	X	X
<i>Leptodactylus podicipinus</i>	2	Least Concern	X	X	X
<i>Hypsiboas punctatus</i>	1	Least Concern	X	X	X
<i>Leptodactylus boliviensis</i>	5	Least Concern	X	X	X
<i>Pipa pipa</i>	1	Least Concern	X	X	X
<i>Amazonphrynela_vote</i>	1	-	X	X	?
<i>Allobates sp1 madeira</i>	15	Least Concern	?	X	X
<i>Hypsiboas sibleszi</i>	1	Least Concern	?	X	X

<i>Chiasmocleis jimi</i>	10	Data Deficient	?	X	?
<i>Hydrolaetare caparu</i>	1	Data Deficient	?	?	X*
<i>Altigius alios</i>	2	Data Deficient	?	?	X
<i>Elachistocleis</i> sp.	3	-	?	?	?
<i>Dendropsophus</i> sp14	2	-	?	?	?
<i>Leptodactylus</i> gr. <i>marmoratus</i>	7	-	?	?	?
<i>Dendropsophus</i> gr. <i>brevifrons</i>	7	-	?	?	?
<i>Scinax</i> sp1	6	-	?	?	?

450

451

452 **3.2 Species distribution models for impact assessments**

453

454 Fourteen species (87.5%) considered to be local-scale targets based on their
 455 distributions in the affected area (group 3) had less than 10 records of occurrence for the
 456 parameterization of the models at this scale. Of these species, the potentially
 457 undescribed species (*Elastichocleis* sp., *Dendropsophus* sp.14, *Leptodactylus* gr.
 458 *marmoratus*, *Dendropsophus* gr. *brevifrons*, and *Scinax* sp1) had the same number of
 459 records at both local and wider spatial scales because they were only detected in the
 460 study site.

461

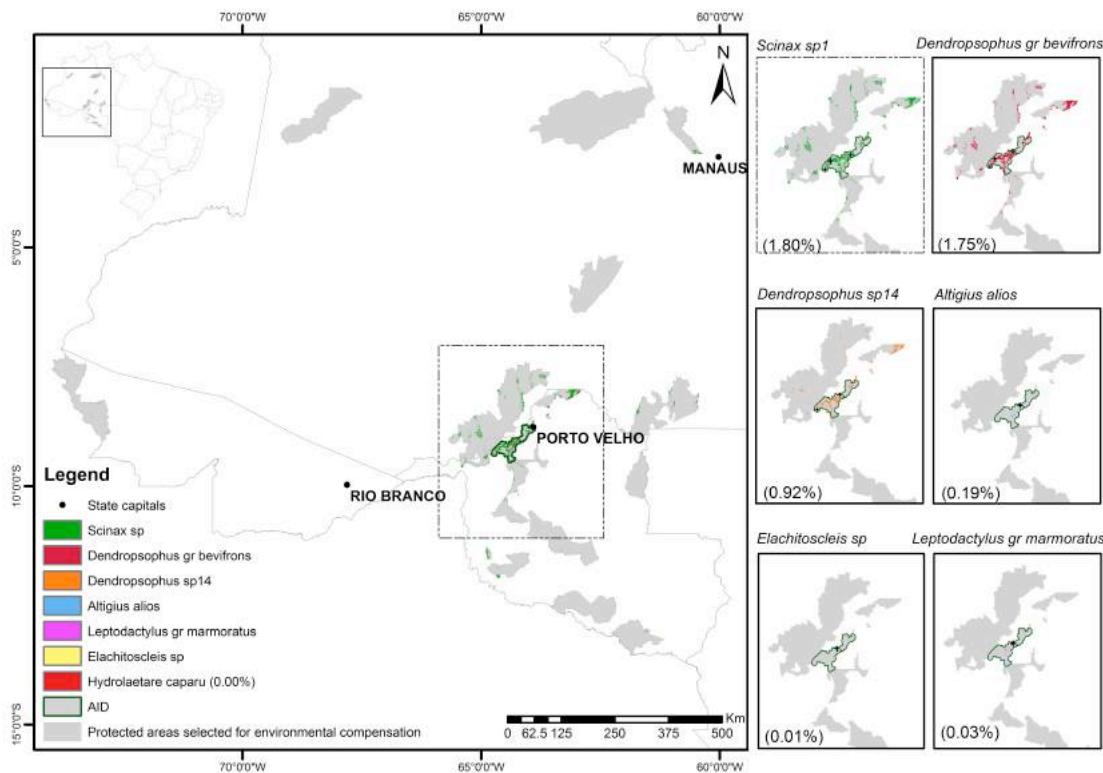
462 **3.2.1 Effects of environmental interpolations and extrapolations on
 463 management decisions**

464

465 Less than 2% of the area in the mosaic of protected units was similar to or
 466 representative of the sampling locations where the target species were detected. Because
 467 over 98% of the pixels selected to receive compensation had at least one variable
 468 outside the data range used to train the models, they were considered novel
 469 environments. Among the 21 reserves, eight did not have areas similar to the occurrence

470 sites of the target species. Moreover, *Hydrolaetare caparu* could not be analyzed
 471 because it only had one occurrence record.

472



473

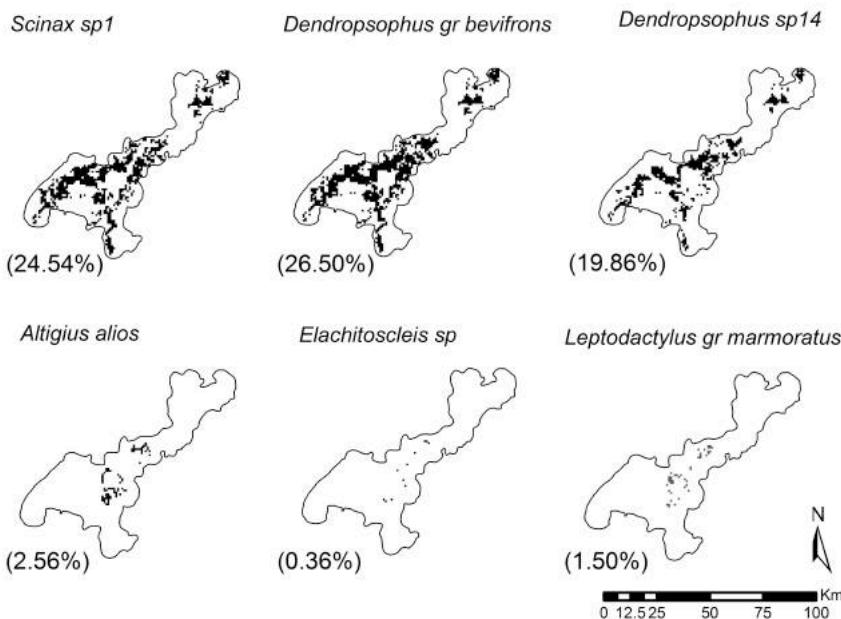
474 **Figure 3:** The results of the multivariate environmental similarity surfaces (MESS) method for
 475 each target species and the mosaic of protected areas programmed to receive compensation
 476 funds. The highlighted areas had environmental values similar to areas in which the individual
 477 species was detected.

478

479

480 At the local scale, similarity values were more representative and ranged from
 481 0.36% for the *Elachistocleis* sp. model to 26.5% for the *Dendropsophus* gr. *bevifrons*
 482 model. MESS maps results were influenced by the complete distribution of the
 483 reference points (Elith et al. 2010, Franklin 2009, Kadmon et al. 2003), which may
 484 explain the low similarity values between *Leptodactylus* gr. *marmoratus* and
 485 *Dendropsophus* gr. *bevifrons* in the MESS model, even though the same number of
 486 records was used for both the testing and training processes.

487



488

489 **Figure 4:** Multivariate environmental similarity surfaces (MESS) for each target species (except
 490 *H. caparu*) in the directly affected area (AID) for use in environmental assessments.
 491
 492

493 **3.2.2 Effects of geographical interpolations and extrapolations on
 494 management decisions**

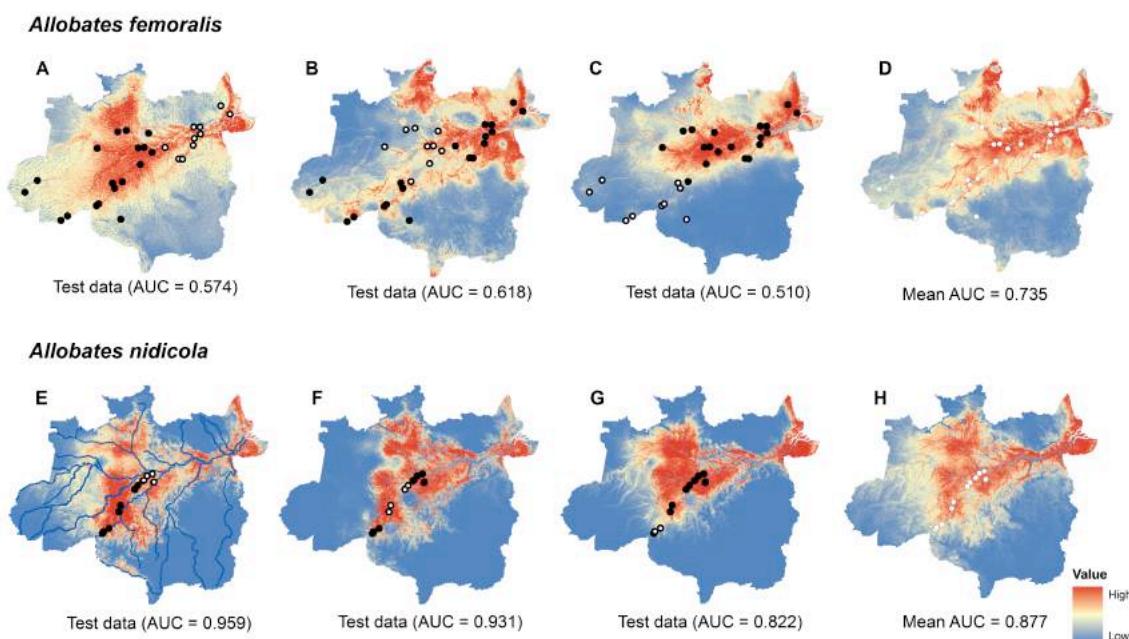
495

496 Models generated with known species distributions at both spatial scales
 497 exhibited lower AUC values (Figure 4-C, D and Figure 3-A, E, F) when the test records
 498 were predicted by the extrapolation of training records. Models in which the validation
 499 records were within the distribution of the training records (interpolation) had better
 500 performances (Figure 5-A, C, G and Figure 6-B, E). The large-scale prediction of *A.*
 501 *nidicola* (Figure 5-E), which was intended to represent the extrapolation arrangement,
 502 exhibited a better AUC performance. However, because of the amount of available data,
 503 the training partition was located adjacent to the training test (interpolation), which may
 504 explain the good performance of the model. At a local scale, the models also had

505 omission errors and low performances; however, there was less variation between the
 506 AUC of the test predictions generated by extrapolation and interpolation of the data
 507 (Figure 6).

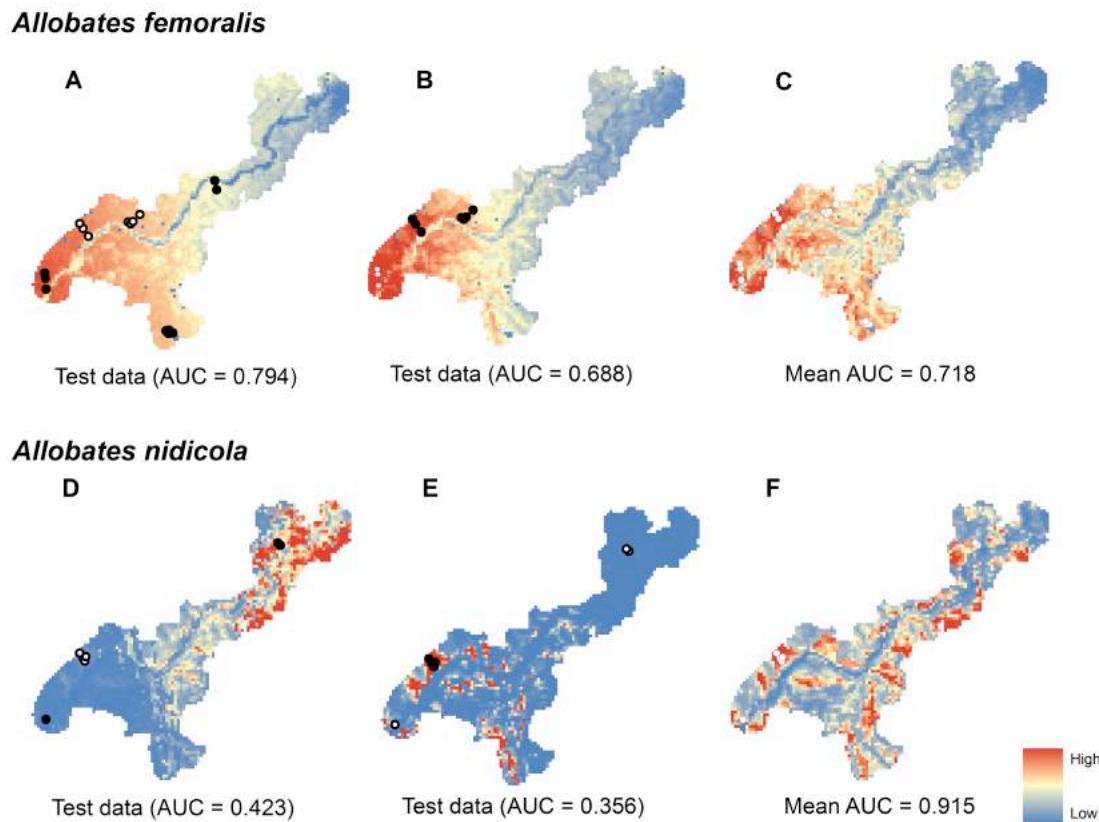
508 Omission and commission errors were found for the biome-scale *A. nidicola*
 509 models because the species was predicted to be widespread in many parts of the basin
 510 even though it is known to be restricted to the Purus-Madeira interfluve. The predictions
 511 indicated greater environmental suitability along the lower reach of Amazon River than
 512 in the region where the taxa actually occur.

513



514

515
 516 **Figure 5:** Predictions of environmentally suitable areas for the occurrence of *Allobates*
 517 *femoralis* and *Allobates nidicola* at the scale of the Brazilian Amazon Basin. Records of
 518 occurrence for each species were separated into 3 groups according to distance for evaluating
 519 the predictions using model interpolation and extrapolation when sampling was incomplete. The
 520 training records for the simulation models are black; the validation points are white.
 521 Models “d” and “h” included all available occurrences; 20% of the occurrences were randomly
 522 assigned by the algorithm to test the models. The AUC values reflected the ability of the model
 523 to predict the test points.
 524



525
 526
 527 **Figure 6:** Predictions of environmentally suitable areas for the occurrence of *Allobates*
 528 *femoralis* and *Allobates nidicola* at the local scale (directly affected area - AID). Because of the
 529 small number of data points at this spatial scale, the records were divided into two groups to
 530 simulate the extrapolated and interpolated predictions. The points in black were used for
 531 training; white dots indicate the test records. Models “C” and “F” included all available
 532 occurrences; 20% of the occurrences were randomly assigned by the algorithm to test the
 533 models.

534
 535

536 **4. Discussion**

537

538 **4.1 Target species selection**

539

540 Some of the criteria suggested in current Brazilian legislation for identifying
 541 target species in environmental impact studies (Resolution CONAMA No. 001/86, IN
 542 IBAMA No. 146/07) do not reflect the type of damage expected in the Amazon Basin;
 543 therefore, these criteria were not suitable for identifying conservation priorities in the

region. The criteria related to the degree of threat for a species according to categories adopted by the IUCN or national lists of threatened species proved particularly uninformative because all of the target species considered in this study were not listed or were categorized as “data deficient”. Furthermore, the degree of threat in the classification is dependent on the spatial scale of the analysis; therefore, many of the taxa categorized as threatened according to national and international lists are not considered threatened in the Amazon biome. For example, the installation of a new development in the Amazon may favor giant otter (*Pteronura brasiliensis*) population growth; the giant otter is a species categorized as endangered (EN) by the IUCN and as vulnerable (VU) on the Brazilian red list (Rosas et al. 2007). Such lists may have national value; however, their use for prioritizing targets should be reviewed and adapted to regional scales.

Because of the substantial knowledge gaps that are characteristic of tropical regions (Bini et al. 2006, Tuomisto and Ruokolainen 1997), and the short time frame for evaluating the impacts on wildlife, target selection should be cautious and grounded in results from studies conducted in the project area. These studies must include sample designs adapted to measure the impact of the installation on elements of local biodiversity in relation to the known distributions of these elements (Magnusson et al. 2013).

In the Santo Antônio case study, most of the species defined as targets have not yet been described in the literature; therefore, scarce data were available for the model parameterizations. Our results are consistent with the conclusions of Cayuela et al. (2009), who indicated that the taxa most in need of conservation measures in tropical

567 regions are those in which the distribution models are less likely to be successfully
568 implemented.

569 Indirect negative damages can affect key species distal to the flooded areas.
570 However, because of the short time during which measures can be proposed and
571 incipient studies made with comparable sampling designs, such damage might not be
572 determined until the later impact stages. Given the rapid expansion of infrastructure
573 projects in forested areas, the data produced in the impact assessments must be
574 integrated to guide future species distribution studies and mitigation decisions at large
575 geographic scales.

576

577 **4.2.1 Effects of environmental interpolations and extrapolations on**
578 **management decisions**

579

580 Because of possible difficulties in detection and the existence of numerous
581 sampling gaps in the Amazon Basin, many of the taxa surveyed and defined as targets
582 in this study might have distributions that extend well beyond the affected area and
583 might not be compromised by the impacts of the project. However, considering the
584 precautionary principle (CDB 1992) and because the Amazon is a region of high
585 endemism (Kaefer et al 2012, Ribas et al. 2007, Wallace, 1852), licensing agents must
586 make at least the minimal effort required to expand the information on these targets
587 before issuing authorizations and take steps to protect potential new occurrence sites of
588 these species.

589 Predictions of environmental suitability also become highly uncertain and
590 undesirable in cases of policy planning for conservation when models are used to

591 predict areas with environmental characteristics that are dissimilar to the area used to
592 calibrate the model (Thuiller et al. 2004, Williams and Jackson 2007). Extrapolations
593 are susceptible to omission errors (Sinclair et al. 2010) because the dataset used for the
594 model parameterization cannot represent all conditions in the extrapolated region.
595 MESS analysis has been used for evaluating the similarity or novelty of environments
596 to indicate where the models are most informed, to guide decisions based on
597 predictions, and to assist in model interpretation (Elith et al. 2010). Furthermore, MESS
598 can be used to identify and reject models with fitted functions that extrapolate in ways
599 that are biologically implausible.

600 The goal of compensation is to permanently or temporarily compensate for the
601 use of natural resources that become unavailable to the environment (Giasson and
602 Carvalho 2012). In a scenario related to the selection of regions for compensation, the
603 models would have to extrapolate predictions beyond the available calibration data and
604 then predict environmentally appropriate areas for the occurrence of the target species.
605 As seen in the models used for the species of interest, 98% of the mosaic area was not
606 representative of the data used for training; therefore, the SDM tools for guiding
607 management decisions produces highly uncertain results and may not be superior to
608 other approaches, such as the use of expert opinions.

609 Tôrres and Vercillo (2012) suggested that distribution modeling tools might
610 improve the targeting requirements of environmental impact studies. Such requirements
611 would be related to the installation of sampling locations; however, the scarce
612 occurrence records of target species are a major limitation. Models are clearly
613 influenced by the number of records used for predictions (Austin et al. 2006, Carroll
614 and Pearson 1998, Hernandez et al. 2006, Kadmon et al. 2003, Pearce and Ferrier 2000,

615 Rushton et al. 2004, Stockwell and Peterson 2002). Thus, the possible approaches using
616 such methods will be restricted to species that have sufficient occurrence records, which
617 were not found for the targets identified in the Santo Antônio case study. Therefore, the
618 use of these models is restricted to MESS analyses or would require using species not
619 that are considered immediately threatened by the project. In this scenario, the models
620 could only assist in other stages of the licensing process (e.g., post-impact monitoring)
621 or as a request for clarification or supplementation to the analyses of the studies (Res
622 CONAMA No. 237/97, IV) because the sample design and resulting sample sites in the
623 environmental studies are predefined in the Terms of Reference document according to
624 secondary biodiversity data for the area that will be impacted.

625 Secondary data are not available for the vast majority of Amazonian regions,
626 which precludes the use of these tools to guide the installation of sampling units in the
627 initial stage of the study prior to the issuance of the license. Nevertheless, additional
628 restrictions, which are discussed in the literature, should be considered prior to analysis.
629 The most important restriction refers to space disagreement between the sampled areas,
630 which are typically defined by licensing agents as the regions adjacent to the affected
631 area (directly affected area - buffer Figure 1), and areas where decisions concerning
632 compensation must be made, which are typically outside that area.

633

634 **4.2.2 Effects of geographical interpolations and extrapolations on**
635 **management decisions**

636

637 The simulations with *A. nidicola* and *A. femoralis* data showed that at the local
638 and biome-wide spatial scales, the models for which geographic predictions were

639 extrapolated beyond the training area had poor predictive performance compared to
640 predictions generated by interpolation. Studies have shown that the projections of
641 models through time and space can be variable (Barbet-Massin et al. 2010, Dormann et
642 al. 2007, Thuiller et al. 2004, Zurell et al. 2012) and are not recommended for the
643 guidance of policy decisions without prior validation.

644 The problems posed by extrapolation are likely to be even greater in the Amazon
645 compared to more topographically and climatically variable areas because species
646 interactions or historical factors that are difficult to quantify are likely to be more
647 important than the environmental data available for modeling (Ribas et al. 2007,
648 Wallace 1852). The BAM diagram (Saupe et al. 2012, Soberón and Peterson 2005,
649 Soberón 2007) illustrates other critical points for the formulation of robust models.
650 When estimates of the area of occurrence are required for compensation or relocation of
651 the project, studies must ascertain favorable conditions for the occurrence of the species
652 (environmental predictors simulating the fundamental ecological niche) and the factors
653 that restrict the species' spread (biotic and historical factors that restrict geographic
654 dispersion). Otherwise, commission errors will be inevitable (Godsoe 2010, Saupe et al.
655 2012), as seen in the models for *A. nidicola* at large spatial scales. The predictions of
656 these models indicated greater environmental suitability of the species near the mouth
657 of the Amazon River than in the region where the taxon is actually present, which is
658 restricted to the Purus-Madeira interfluve.

659 Despite the classic studies of biogeographical processes that influence the
660 distribution patterns of organisms in Amazonia (Gascon et al. 1998, Gascon et al. 2000,
661 Kaefer et al. 2012, Ribas et al. 2007, Ribas et al. 2011, Wallace 1852), many of these
662 processes remain unknown. Moreover, there is even disagreement about the most

663 important historical events that occurred in the region (e.g., Hoorn et al. 2010, Rossetti
664 et al. 2005). Such historical and geographical processes are independent of the current
665 environmental conditions; even with the inclusion of information regarding past
666 climates and paleolandscapes, directed studies with targets are necessary to evaluate
667 other possible dispersal barriers.

668 With the appropriate precautions (as cited above), SDMs could be used to guide
669 new collection sites and clarify unanswered questions in previous studies . In the case
670 presented here, lawmakers were seeking potential environmentally appropriate areas
671 within the geographic occurrence of the target taxa (Elith et al. 2009). However, the
672 processes that shape biodiversity vary with the spatial scale of the model; therefore, the
673 use of inappropriate scales can compromise the usefulness of the models and
674 misrepresent the distribution patterns (Guisan and Thuiller 2005, Hortal et al. 2010,
675 Kristiansen et al. 2009, Nogués-Bravo and Araújo 2006, Whittaker et al. 2001).

676 Additional sample units that both record species occurrence and measure environmental
677 predictors that cannot be obtained by remote sensing are required. However, the
678 extrapolation of these variables to regions beyond the area of origin involves many
679 uncertainties and is dependent on the nature of the predictor.

680 Despite the present adverse scenario, species distribution models have great
681 potential to assist in conservation by contributing to strategic decisions regarding
682 environmental impacts in tropical regions (Cayuela et al. 2009). For this potential to be
683 realized, environmental agencies must develop strategies to reduce spatial uncertainties
684 in predictions of species distribution while also investing in improvements to the current
685 state of knowledge of the region's biodiversity. An alternative for overcoming the
686 challenges listed here might be to expand the sample area defined by the licensing

687 agents beyond the regions proximal to the project's affected area. If this action is
688 performed, the use of permanent plots with multiple surveys could provide data on the
689 detection probabilities for better evaluating the probability of false absences.
690 Furthermore, sampling larger areas would allow the interpolation of collected data and
691 reduce inaccuracies caused by geographical and environmental extrapolation models.
692 Refined environmental variables could also be collected in the field to produce better
693 local-scale models. If the sample expansion conflicts with urgent deadlines for installing
694 economic ventures, an alternative conservation strategy would be to invest in
695 consolidate programs of long-term ecological research that are integrated with surveys
696 to monitor the biodiversity in the region. These surveys should be performed before
697 short-term studies that are associated with licensing individual sites are required (Costa
698 and Magnusson 2010, Magnusson et al. 2005, Magnusson et al. 2013).
699

700 **5. Conclusions**

701

702 We conclude that the lack of knowledge regarding the distribution of species
703 poses a risk to biodiversity when potentially damaging enterprises are licensed in the
704 Amazon. However, the use of modeling tools to overcome such deficits and to guide the
705 requirements within impact assessments remains challenging because of the large areas
706 involved in management decisions, limited sampling associated with impacts, and the
707 lack of secondary data for calibration of the models in the region. SDMs may be useful
708 for some aspects of licensing processes in the Amazon; however, managers must
709 understand the practical limitations of these tools for answering the demands identified
710 by Tôrres and Vercillo (2012).

711 Because of the urgency of obtaining information related to species distributions
712 in the Amazon, the responsibility of filling these gaps in biological information cannot
713 rest exclusively with the scientific community. Many projects are being planned for the
714 Amazon region; therefore, it is important that the information required by the
715 environmental licensing procedures in the Amazon be comparable to perform analyses
716 of biological complementarity between regions in the basin on short notice and to
717 enable the future use of more realistic models with robust regional information.

718

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720

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731

732 **7. References**

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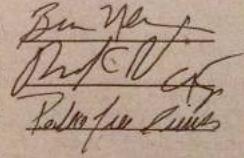
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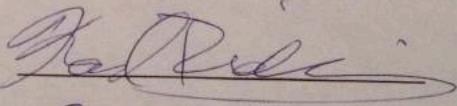
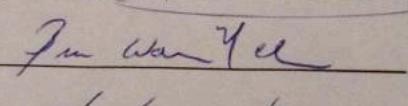
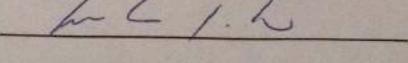
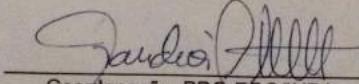
8. Attachments

8.1. ATAS DAS BANCAS AVALIADORAS

8.1.1. Ata da banca avaliadora da aula de qualificação:

			
AULA DE QUALIFICAÇÃO			
PARECER			
Aluno(a): LORENA RIBEIRO DE ALMEIDA CARNEIRO Curso: ECOLOGIA Nível: MESTRADO Orientador(a): WILLIAM ERNEST MAGNUSSON			
Título: <i>"A modelagem preditiva de distribuição de espécies e o desafio de sua utilização em processos de licenciamento ambiental de hidrelétricas na Amazônia"</i>			
BANCA JULGADORA:			
TITULARES: Bruce Walker Nelson (INPA) Richard Carl Vogt (INPA) Pedro Ivo Simões (INPA)		SUPLENTES: Camila Cherem Ribas (INPA) Maria Ermelinda Oliveira (UFAM)	
PARECER		ASSINATURA	
Bruce Walker Nelson (INPA) <input checked="" type="checkbox"/> Aprovado <input type="checkbox"/> Reprovado Richard Carl Vogt (INPA) <input checked="" type="checkbox"/> Aprovado <input type="checkbox"/> Reprovado Pedro Ivo Simões (UFAM) <input checked="" type="checkbox"/> Aprovado <input type="checkbox"/> Reprovado Camila Cherem Ribas (INPA) <input type="checkbox"/> Aprovado <input type="checkbox"/> Reprovado Maria Ermelinda Oliveira (UFAM) <input type="checkbox"/> Aprovado <input type="checkbox"/> Reprovado			
Manaus(AM), 11 de abril de 2012			
OBS: <hr/> <hr/> <hr/> <hr/>			
<small>INSTITUTO NACIONAL DE PESQUISAS DA AMAZÔNIA INPA PROGRAMA DE PÓS-GRADUAÇÃO EM ECOLOGIA PPG-ECO Av. Elígenio Sales, 2239 – Bairro: Aleixo – Caixa Postal: 478 – CEP: 69.011-970, Manaus/AM. Fone: (+ 55) 92 3643-1909 Fax: (+ 55) 92 3643-1909 e-mail: mcacolonia@gmail.com</small>			

8.1.2. Ata da banca avaliadora da defesa de mestrado:

 INPA <small>INSTITUTO NACIONAL DE PESQUISAS DA AMAZÔNIA</small>	 PG-ECO-INPA <small>PÓS-GRADUAÇÃO EM ECOLOGIA</small>	<small>GOVERNO FEDERAL</small>  BRASIL <small>PAÍS RICO É PAÍS SEM Pobreza</small>
ATA DA DEFESA PÚBLICA DA DISSERTAÇÃO DE MESTRADO DO PROGRAMA DE PÓS-GRADUAÇÃO EM ECOLOGIA DO INSTITUTO NACIONAL DE PESQUISAS DA AMAZÔNIA.		
<p>Aos 12 dias do mês de setembro do ano de 2013, às 14:30 horas, na sala de aula do Programa de Pós Graduação em Ecologia – PPG ECO /INPA, reuniu-se a Comissão Examinadora de Defesa Pública, composta pelos seguintes membros: o(a) Prof(a). Dr(a). Karl Didier, da Wildlife Conservation Society - WCS, o(a) Prof(a). Dr(a). Bruce Walker Nelson, do Instituto Nacional de Pesquisas da Amazônia – INPA e o(a) Prof(a). Dr(a). Marcelo Gordo, da Universidade Federal do Amazonas - UFAM, tendo como suplentes o(a) Prof(a). Dr(a). Marina Anciães, do Instituto Nacional de Pesquisas da Amazônia – INPA e o(a) Prof(a). Dr(a). Fabricio Beggiato Baccaro, da Universidade Federal do Amazonas - UFAM, sob a presidência do(a) primeiro(a), a fim de proceder a arguição pública do trabalho de DISSERTAÇÃO DE MESTRADO DE LORENA RIBEIRO DE ALMEIDA CARNEIRO, intitulado "Limitações no uso da modelagem de distribuição de espécies no contexto das avaliações de impacto ambiental de empreendimentos na Amazônia", orientado pelo(a) Prof(a). Dr(a). William Ernest Magnusson, do Instituto Nacional de Pesquisas da Amazônia – INPA e coorientado(a) pelo(a) Prof(a). Dr(a). Albertina Lima, do Instituto Nacional de Pesquisas da Amazônia – INPA e pelo(a) Prof(a). Dr(a) Ricardo Bonfim Machado, da Universidade de Brasília– UnB.</p>		
<p>Após a exposição, o(a) discente foi arguido(a) oralmente pelos membros da Comissão Examinadora, tendo recebido o conceito final:</p>		
<input checked="" type="checkbox"/> APROVADO(A)	<input type="checkbox"/> REPROVADO(A)	
<input checked="" type="checkbox"/> POR UNANIMIDADE	<input type="checkbox"/> POR MAIORIA	
<p>Nada mais havendo, foi lavrada a presente ata, que, após lida e aprovada, foi assinada pelos membros da Comissão Examinadora.</p>		
<small>Prof(a).Dr(a). Karl Didier</small>		
<small>Prof(a).Dr(a). Bruce Walker Nelson</small>		
<small>Prof(a).Dr(a). Marcelo Gordo</small>		
 <small>Coordenação PPG-ECO/INPA</small>		

8.1.3. Ata da banca avaliadora da dissertação (trabalho escrito):



Instituto Nacional de Pesquisas da Amazônia - INPA
Programa de Pós-graduação em Ecologia



Avaliação de dissertação de mestrado

Título: **Limitações no uso da modelagem preditiva de distribuições de espécies em avaliações de impacto ambiental de empreendimentos na Amazônia**

Aluno: **LORENA RIBEIRO DE ALMEIDA CARNEIRO**

Orientador: **William Magnusson** Co-orientadores: **Ricardo Machado e Albertina Lima**

Avaliador: Eduardo Venticinque

Por favor, marque a alternativa que considerar mais apropriada para cada ítem abaixo, e marque seu parecer final no quadro abaixo

	Muito bom	Bom	Necessita revisão	Reprovado
Relevância do estudo	(x)	()	()	()
Revisão bibliográfica	()	(x)	()	()
Desenho amostral/experimental	()	(x)	()	()
Metodologia	()	()	(x)	()
Resultados	()	()	(x)	()
Discussão e conclusões	()	()	(x)	()
Formatação e estilo texto	()	(x)	()	()
Potencial para publicação em periódico(s) indexado(s)	()	(x)	()	()

PARECER FINAL

Aprovada (indica que o avaliador aprova o trabalho sem correções ou com correções mínimas)

Aprovada com correções (indica que o avaliador aprova o trabalho com correções extensas, mas que não precisa retornar ao avaliador para reavaliação)

Necessita revisão (indica que há necessidade de reformulação do trabalho e que o avaliador quer reavaliar a nova versão antes de emitir uma decisão final)

Reprovada (indica que o trabalho não é adequado, nem com modificações substanciais)

Lima, 27/04/2013,
 Local Data

Assinatura

Comentários e sugestões podem ser enviados como uma continuação desta ficha, como arquivo separado ou como anotações no texto impresso ou digital da tese. Por favor, envie a ficha assinada, bem como a cópia anotada da tese e/ou arquivo de comentários por e-mail para pgecologia@gmail.com e flaviacosta001@gmail.com ou por correio ao endereço abaixo. O envio por e-mail é preferível ao envio por correio. Uma cópia digital de sua assinatura será válida.

Endereço para envio de correspondência:

Flavia Costa
 DCEC/CPEC/INPA
 CP 478
 69011-970 Manaus AM
 Brazil



Instituto Nacional de Pesquisas da Amazônia - INPA
Programa de Pós-graduação em Ecologia



Avaliação de dissertação de mestrado

Título: Limitações no uso da modelagem preditiva de distribuições de espécies em avaliações de impacto ambiental de empreendimentos na Amazônia
 Aluno: LORENA RIBEIRO DE ALMEIDA CARNEIRO

Orientador: William Magnusson Co-orientadores: Ricardo Machado e Albertina Lima

A. Townsend Peterson

Por favor, marque a alternativa que considerar mais apropriada para cada ítem abaixo, e marque seu parecer final no quadro abaixo

	Muito bom	Bom	Necessita revisão	Reprovado
Relevância do estudo	(X)	()	()	()
Revisão bibliográfica	()	()	(X)	()
Desenho amostral/experimental	()	()	(X)	()
Metodologia	()	()	(X)	()
Resultados	()	()	(X)	()
Discussão e conclusões	()	(X)	()	()
Formatação e estilo texto	(X)	()	()	()
Potencial para publicação em periódico(s) indexado(s)	()	()	(X)	()

PARECER FINAL

Aprovada (indica que o avaliador aprova o trabalho sem correções ou com correções mínimas)

Aprovada com correções (indica que o avaliador aprova o trabalho com correções extensas, mas que não precisa retornar ao avaliador para reavaliação)

Necessita revisão (indica que há necessidade de reformulação do trabalho e que o avaliador quer reavaliar a nova versão antes de emitir uma decisão final)

Reprovada (indica que o trabalho não é adequado, nem com modificações substanciais)

Lawrence, Kansas, USA
 Local

5 May 2013,
 Data

Assinatura

Comentários e sugestões podem ser enviados como uma continuação desta ficha, como arquivo separado ou como anotações no texto impresso ou digital da tese. Por favor, envie a ficha assinada, bem como a cópia anotada da tese e/ou arquivo de comentários por e-mail para pgecologia@gmail.com e flaviacosta001@gmail.com ou por correio ao endereço abaixo. O envio por e-mail é preferível ao envio por correio. Uma cópia digital de sua assinatura será válida.

Endereço para envio de correspondência:

Flavia Costa
 DCEC/CPEC/INPA
 CP 478
 69011-970 Manaus AM
 Brazil



Avaliação de dissertação de mestrado

Título: Limitações no uso da modelagem preditiva de distribuições de espécies em avaliações de impacto ambiental de empreendimentos na Amazônia
 Aluno: LORENA RIBEIRO DE ALMEIDA CARNEIRO

Orientador: William Magnusson Co-orientadores: Ricardo Machado e Albertina Lima

Avaliador: José alexandre Felizola Diniz Filho

Por favor, marque a alternativa que considerar mais adequada para cada ítem abaixo, e marque seu parecer final no quadro abaixo

	Muito bom	Bom	Necessita revisão	Reprovado
Relevância do estudo	(X)	()	()	()
Revisão bibliográfica	()	(X)	()	()
Desenho amostral/experimental	()	(X)	()	()
Metodologia	()	(X)	()	()
Resultados	(X)	()	()	()
Discussão e conclusões	(X)	()	()	()
Formatação e estilo texto	(X)	()	()	()
Potencial para publicação em periódico(s) indexado(s)	()	(X)	()	()

PARECER FINAL

(X) **Aprovada** (indica que o avaliador aprova o trabalho sem correções ou com correções mínimas)

() **Aprovada com correções** (indica que o avaliador aprova o trabalho com correções extensas, mas que não precisa retornar ao avaliador para reavaliação)

() **Necessita revisão** (indica que há necessidade de reformulação do trabalho e que o avaliador quer reavaliar a nova versão antes de emitir uma decisão final)

() **Reprovada** (indica que o trabalho não é adequado, nem com modificações substanciais)

Goiânia, GO
 Local

03/05/2013
 Data

Assinatura

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Endereço para envio de correspondência:

Flavia Costa
 DCEC/CPEC/INPA
 CP 478
 69011-970 Manaus AM