PHILOSOPHICAL TRANSACTIONS OF THE ROYAL SOCIETY B

BIOLOGICAL SCIENCES

Estimating Interaction Credit for Trophic Rewilding in Tropical Forests

Journal:	Philosophical Transactions B
Manuscript ID	RSTB-2017-0435.R1
Article Type:	Research
Date Submitted by the Author:	20-Aug-2018
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Issue Code (this should have already been entered but please contact the Editorial Office if it is not present):	REWILD
Subject:	Ecology < BIOLOGY
Keywords:	Atlantic Forest, defaunation, seed dispersal, plant-animal interaction, reintroduction, restoration

SCHOLARONE[™] Manuscripts

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22 Abstract

Trophic rewilding has been suggested as a restoration tool to reverse defaunation and its cascading effects on ecosystem functioning. One of the ecological processes that has been jeopardized by defaunation is animal-mediated seed dispersal. Here, we propose an approach that combines joint species distribution models with occurrence data and species interaction records to quantify the potential to restore seed-dispersal interactions through rewilding and apply it to the Atlantic Forest, a global biodiversity hotspot. Using this approach, we identify areas that should benefit the most from trophic rewilding and candidate species that could contribute to cash the credit of seed-dispersal interactions in a given site. We found that sites within large fragments bearing a great diversity of trees may have about 20 times as many interactions to be cashed through rewilding as small fragments in regions where deforestation has been pervasive. We also ranked mammal and bird species according to their potential to restore seed-dispersal interactions if reintroduced while considering the biome as a whole and at finer scales. Our findings can aid future conservation efforts in rewilding projects in defaunated tropical rainforests.

39 Keywords: Atlantic Forest, defaunation, seed dispersal, plant-animal interaction,

40 reintroduction, restoration

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41	1. Introduction
42	The pervasive biodiversity crisis we live in has prompted active conservation
43	approaches to reverse the effects of defaunation [1,2]. As animal populations and
44	species decline in natural environments, the ecological interactions involving them are
45	also lost, threatening the functioning of ecological systems [3]. Trophic rewilding,
46	defined as species reintroductions and surrogate introductions to restore ecological
47	interactions [4,5], is increasingly considered as one of the few viable options to reinstate
48	ecosystem functions [6]. However, species introductions entail intensive planning
49	integrating detailed natural history and ecological knowledge to ensure the desired
50	results while reducing the potential risks [4,7].
51	Although rewilding has become an important debate in the last years [8,9], few
52	projects in the field have applied rewilding for mitigating the loss of species interactions
53	[10,11]. Because rewilding focuses on restoring ecological processes [4,9], the choice of
54	the candidate species should be based on the balance between the probability of
55	population establishment and the benefit of the restored interactions for ecosystem
56	functions [4,5]. Using an economics analogy, species reintroductions can allow
57	defaunated areas to gradually "cash" a credit of ecological interactions where their
58	interaction partners are still extant, which depends on the species-specific traits as well
59	as on the abundances of the focal species and their partners [12]. The interaction credit
60	framework is based on the potential to restore interactions (i.e. to reconnect species that
61	became disconnected) in an area following reintroduction; thus, it can be used to predict
62	and to evaluate the success of rewilding in reinstating ecological processes.
63	Restoring certain types of interactions, such as seed dispersal, can be particularly
64	beneficial because it helps natural forest regeneration and may even aid the restoration

64 beneficial because it helps natural forest regeneration and may even aid the restoration

of neighbouring areas, creating more suitable habitat and generating a positive feedback

66	for conservation efforts [13]. Most tropical tree species rely on animal-mediated seed
67	dispersal for recruitment [14]. Yet, many frugivore species that establish non-redundant
68	seed-dispersal interactions tend to be the same that are overhunted and affected by
69	fragmentation [15,16]. Therefore, local extinctions can cause the loss of those
70	ecological interactions, cascading to the loss of functioning in defaunated tropical areas
71	[17].
72	Tropical forests are highly threatened worldwide but the Atlantic Forest in South
73	America represents one of the worst-case scenarios for protecting biodiversity: it has
74	been severely defaunated and is highly fragmented [18,19], with many of the fragments
75	too small to maintain viable populations of certain species. Moreover, the agricultural
76	and urban matrices surrounding forest patches prevent some animals from recolonizing
77	patches where they have been extirpated [20]. Thus, the reestablishment of ecological
78	interactions even in large fragments is unlikely to occur without more active
79	approaches, such as trophic rewilding [7,21].
80	Despite the large amount of data available on species distributions and local
81	interaction patterns, predicting how the reintroduced species will interact with the local
82	community is still an obstacle [12,22,23]. Devising a comprehensive framework for
83	trophic rewilding requires integrating information on species distributions and
84	interaction patterns across their range to allow inferring which interactions are likely or
85	unlikely to occur [23]. A spatially informed framework is needed to aid decision-
86	making regarding the prioritization of the areas and the choice of candidate species for
87	trophic rewilding at broad scales [24].
88	Here we use a probabilistic framework that combines joint species distribution
89	models [25], the ecological network approach [23] and the credit of ecological
90	interactions framework [12] to identify priority areas and candidate species for trophic

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91	rewilding, with a focus on restoring seed-dispersal interactions. We use the Atlantic
92	Forest as a case study and show how the proposed framework can help to unveil
93	interaction credit hotspots and to detect which species are the best candidates for
94	rewilding at different spatial scales.
95	
96	2. Methods
97	The credit of ecological interactions corresponds to the number of animal-plant
98	interactions expected to be restored if an extirpated species is reintegrated into a given
99	area [12]. Therefore, quantifying interaction credit requires a toolset for predicting
100	pairwise species interactions in a locality. The prerequisite for an interaction to take
101	place is that the focal species co-occur in space and time. Although species occurrences
102	can be inferred directly from occurrence data, modelling species occurrences, besides
103	allowing predictions for poorly sampled locations, offers the possibility of
104	encompassing uncertainty regarding occurrence and co-occurrence patterns. Species
105	may co-occur because they respond in the same manner to the environment, but co-
106	occurring does not mean species will necessarily interact. Assuming that two species
107	co-occur in space and time, interactions will still depend on a number of factors, such as
108	species traits, phenology and abundance [12,26,27]. Because of the inherent challenges
109	of detecting and predicting interactions in a community, the appropriate way to model
110	interaction patterns is using a probabilistic approach [23,28]. Pairwise interaction
111	probabilities can be obtained by the element-wise product between the matrix O ,
112	depicting species co-occurrence probabilities and matrix A, depicting the expected
113	interaction probabilities once the species co-occur [23]. Considering that the aim of
114	trophic rewilding is to restore processes that have been lost over time, its outcomes in
115	terms of interaction credit can be predicted and then evaluated by comparing expected

116	interactions under two different scenarios: (1) a benchmark historical scenario, and (2)
117	the current scenario where many sites may be defaunated relative to the historical
118	benchmark. We apply the framework outlined above to estimate the credit of ecological
119	interactions across the Atlantic Forest in South America.
120	
121	(a) Data
122	To infer current distribution patterns of frugivores and plants, we used the most
123	spatially and taxonomically comprehensive community databases available for the
124	Brazilian Atlantic Forest (the ATLANTIC series data papers:
125	https://github.com/LEEClab/Atlantic_series; and the Neotropical Tree Communities
126	database TreeCo version 2.0: <u>http://labtrop.ib.usp.br/doku.php?id=projetos:treeco:start</u>).
127	We used data on all seed disperser genera with interactions recorded in the Atlantic
128	Forest [29], excluding species known to behave more as seed predators than dispersers
129	as well as bats and small mammals due to incomplete data on distribution or interaction
130	patterns (which could impair the assessment of a spatial credit at broad scale) and to
131	taxonomic inconsistency in some genera (which could impair the prediction of pairwise
132	interactions). Therefore, we compiled data on 211 vertebrate species (birds and medium
133	to large-sized mammals) and 1,426 tree and palm species [30-33]. See electronic
134	supplementary material, appendix A, for the filtered reference list of the plant
135	occurrence data. We only considered plant species that were identified as zoochoric
136	[29,34], reported to grow \geq 4 meters high, and having \geq 5 observations in the plant
137	occurrence dataset. For the full list of included species, see electronic supplementary
138	material, appendix B. To guarantee a broad coverage across the entire biome, we
139	created a regular grid of 40,000 prediction sites spanning the Atlantic Forest in ArcGIS
140	software (version 10.3) and selected for later use those 912 sites that overlapped with

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141 Atlantic Forest fragments. To assess the interaction component, we used interaction

142 records from the ATLANTIC-FRUGIVORY dataset [29]. From this dataset we built the

143 final interaction matrix **P**, as described in more detail below.

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145 (b) Co-occurrence probabilities

We inferred co-occurrence probabilities for frugivores and plants across the Atlantic 146 Forest using joint species distribution models (HMSC [25]). HMSC helps to overcome 147 148 the problem of low spatial overlap in the original surveys across taxonomic groups, generating predicted occurrence probabilities for each frugivore and plant species across 149 150 the Atlantic Forest (for more detailed description on the model construction and included data, see electronic supplementary material, appendix C). The predicted 151 probabilities of species occurring in a given site are determined not only by the values 152 of environmental covariates on that site (through the fixed effect part of the HMSC), but 153 also on the occurrences of the focal and other species in nearby sampling sites (through 154 spatially structured latent variables included in HMSC, see [35]). We generated 500 155 predicted communities in each of the 912 prediction sites described above. For each 156 157 prediction site, we sampled model parameters from the posterior distribution, and thus the predictions account for parameter uncertainty. 158

To generate a historical benchmark distribution of frugivores, we used the distribution records of frugivores from IUCN [36] complemented by records from Wikiaves [37] to determine the sub-regions where each species occurs. The Atlantic Forest comprises seven biogeographical sub-regions that differ in environmental variables and biodiversity composition: Interior, Araucaria, Serra do Mar, Bahia, Diamantina, Sao Francisco, and Pernambuco [38]. We assumed in the benchmark scenario that species would have been present in all prediction sites within all sub-

166	regions where the species is known to currently occur, and assigned $p = 1$ for the
167	probability of species occurrence in those sites. We also performed a sensitivity analysis
168	by setting the occurrence probabilities of frugivores in the benchmark scenario equal to
169	the maximum of the mean posterior occurrence probability across all prediction sites.
170	Because our main interest was in the effects of reintroducing the frugivores, we used the
171	estimated occurrence probabilities for the plants in both the current and the benchmark
172	scenario. Even though it is plausible that certain plants had greater occurrence
173	probabilities in certain sites in the past, this would only increase our credit estimates.
174	Thus, by setting the benchmark using the current probabilities for plants we adopt a
175	conservative approach and we can assign differences between the two scenarios to the
176	differences in frugivore occurrence probabilities.
177	We constructed predicted co-occurrence matrices for each site as the product of
178	occurrence probabilities for frugivores (either assuming the benchmark distribution or
179	the current predicted distribution) and plants (assuming the current predicted
180	distribution).
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182	(c) Interaction probabilities
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191	with any plant species within the focal plant species' genus. We assign unlikely
192	interactions a value greater than zero, because differentiating between true and
193	sampling-induced zeros is challenging [39,40] and it is conceptually wrong to assume
194	that unobserved interactions are impossible to occur. To test for the sensitivity of the
195	results to the probabilities assigned to unlikely interactions, we reran all analyses under
196	two additional scenarios: assuming $a_{ij} = 0.05$ and $a_{ij} = 0.001$ for unlikely interactions.
197	
198	(d) Calculating the spatial credit of ecological interactions
199	To estimate the credit of ecological interactions we compared the historical benchmark
200	and the current scenario of seed-dispersal interactions across the Atlantic Forest. First,
201	we obtained the probabilistic interaction network (matrix P , with elements p_{ij}) for each
202	scenario in each prediction site as the element-wise product of the co-occurrence (\mathbf{O})
203	and interaction matrices (A). We then calculated for each prediction site the expected
204	number of interactions, \hat{L} , as the sum of all pairwise interaction probabilities [28]
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205 206 207 208 209 210 211 211	$\hat{L} = \sum p_{ij}$ and computed the site-specific spatial credit of ecological interactions as the difference between \hat{L} for the benchmark and current scenarios. (e) Calculating frugivore-specific contributions to the credit of ecological interactions To understand which species show potential for cashing the highest credit of ecological interactions through rewilding locally and regionally in the Atlantic Forest, we first computed for each site the expected number of interactions of each frugivore <i>i</i> as the

then built a distribution of $\Delta \hat{k}$ for each species considering all prediction sites and ranked the species based on the mean $\Delta \hat{k}$. To compute $\Delta \hat{k}$ we only consider the prediction sites where a species is expected to occur according to the benchmark scenario. We selected 12 species with highest regional scores to explore more closely their potential for restoring seed-dispersal interactions through rewilding. We examined the distribution of their local scores and discussed the ecology, current threats and rewilding feasibility considering these species. We also calculated the regional scores for each biogeographical sub-region separately to illustrate how this approach could help to guide rewilding at a finer scale.

3. Results

The spatial credit of ecological interactions ranged between 209 and 4,814 interactions that could be cashed in prediction sites across the Atlantic Forest (figure 1). Sites with greater credit have about 20 times as many interactions to be cashed through rewilding as the sites with lower credit. The credit was highest in the Interior sub-region and lowest in the Pernambuco sub-region (figure 1; electronic supplementary material appendix D, figure S1). The interaction credit was highest in areas where occurrence probabilities were high for plant species and low for frugivore species as compared to their expected distributions in the benchmark scenario (electronic supplementary material, figures S2; S3). The areas with low interaction credit were either areas having a high frugivore species richness that matches the expectation in the benchmark scenario, or areas where plant species richness is currently low and hence there are less plant species missing frugivore partners. By estimating the expected number of interactions to be restored by each

240 frugivore, we observed a large variation in the credit that could be cashed by different

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241 species, both at local and regional scales. Altogether 13 species scored highest in at least 242 one of the 912 prediction sites, three of which were primates and ten were birds. Species 243 with high local scores were those that are expected to have a large number of 244 interactions, but had been extirpated in many prediction sites. We selected 12 species with the highest regional scores in restoring seed 245 246 dispersal interactions in the Atlantic Forest to examine their local variation in 247 interaction credit more carefully (table 1, figure 2). Even though the southern muriqui, 248 Brachyteles arachnoides, scored highest because of the high average in the potential to contribute with interactions, the credit contribution of the species would be limited due 249 250 to its restricted distribution. The rusty-margined guan, *Penelope superciliaris*, had, in addition to its high score, the lowest variation in local scores, which reflects its broader 251 range encompassing most of the biome. Sixty-six percent of the species with the highest 252 253 regional scores are endemic to the Atlantic Forest. The sub-regions with the largest potential for cashing in the credit of ecological 254 interactions had similar lists of the highest scoring species; for instance, two thirds of 255 256 the highest scoring species were the same for the Interior and Serra do Mar sub-regions 257 (electronic supplementary material appendix D, table S3). The exceptions are those species that are endemic to a certain sub-region and thus may be regionally important 258 259 but are replaced by other species in the sub-regions where they are not known to occur. 260 Using the maximum of the mean posterior occurrence probabilities to build the 261 benchmark scenario (see Methods) yielded similar results, suggesting that our findings 262 are robust to our choice on how to generate a benchmark for comparison (see electronic supplementary material, figure S4 and table S4 in appendix D). In addition, the results 263 264 showed to be robust to our choice for the probability assigned to unlikely interactions (see electronic supplementary material, table S5 and figure S5 in appendix D). 265

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	267	4. Discussion
	268	Combining joint distribution modelling and the network approach under a probabilistic
	269	framework allowed us to identify hotspots in the Atlantic Forest where trophic
	270	rewilding could be most beneficial to promote the restoration of seed-dispersal
	271	interactions. We found high variability in the interaction credit throughout the biome.
	272	Hotspots of interaction credit were more pervasive in the Interior sub-region of the
	273	Atlantic Forest. This sub-region is characterized by a highly fragmented landscape
	274	resulting from the expansion of agricultural activities [19]. Although such land use
	275	changes produced a major loss in forest cover and shrinkage of fragments [19], the high
	276	interaction credit revealed by our analyses suggests that animal-dispersed plants are
	277	expected to occur in several of those fragments, whereas their frugivore partners are
	278	likely to be missing. The decline in seed disperser richness implies that regeneration and
	279	the persistence of plant populations in fragments may be impaired in the long term due
	280	to seed dispersal limitation if the fragments are not rewilded [41], which is an additional
	281	threat to the remaining fragments [42,43]. This highlights an insidious feature of
	282	defaunation: defaunation and its ecological consequences are much harder to detect than
i	283	deforestation [44].
	284	We also identified hotspots for rewilding within some sub-regions known to bear
	285	relatively well-preserved stretches of habitat, such as the Serra do Mar sub-region.
	286	These hotpots are concentrated around the largest urban areas within the biome, which
i	287	are known to have low densities of many animal species, especially large-sized
2	288	frugivores [45]. However, the high richness of plant species due to preserved forest
	289	patches combined with the low occurrence probability of frugivores due to past or
	290	current threats, promote high potential for interaction restoration through rewilding. The

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hotspots detected within sub-regions can be considered promising areas to focus
rewilding strategies on. This reveals that despite the coarse-grained approximations
used here, the proposed framework can help identifying local hotspots for trophic
rewilding.

The low credit of ecological interactions in the northern Atlantic Forest, in 295 296 particular in the Pernambuco sub-region, may be explained by the long history of 297 extensive deforestation in the region [46]. Deforestation results in low occurrence 298 probabilities of many plant species and thus low potential for seed dispersal interactions to be re-established solely by reintroducing frugivores. The example of the northern 299 Atlantic Forest shows that a closer examination of our framework's outputs may also 300 help to inform when trophic rewilding is not the best option due to habitat unsuitability. 301 Low credit may indicate that few interactions remain to be restored in the area. 302 However, studying the patterns of plant species richness and composition within regions 303 304 may shed light on the other underlying causes for the low interaction credit. When the spatial credit of ecological interactions is low mainly due to low plant species richness, 305 rewilding alone will have a minimum impact, and reestablishment the populations of 306 307 animal-dispersed plants would be required prior to fauna reintroductions [47]. Therefore, our approach may also be helpful in identifying among a set of sites those in 308 309 which rewilding alone would not be the most suitable conservation strategy. 310 We found certain frugivores to contribute disproportionately in cashing the 311 credit of seed dispersal interactions in the Atlantic Forest. These are often species with 312 high degree of frugivory that interact with multiple plant taxa and have low probability of occurrence in sites where their plant partners are likely to occur. This list includes 313 314 species threatened with local extinction in many areas, such as the primates *Brachyteles* spp. and the piping guan Aburria jacutinga. However, although these species may be 315

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316	important locally, they have restricted distributions, and their potential as rewilding
317	candidates does not apply to all sub-regions. On the other hand, some species
318	considered common, such as small generalist birds, had high scores throughout the
319	whole biome. These frugivores tend to establish seed-dispersal interactions with many
320	different species, which makes them good candidates despite the fact that they are likely
321	to occur in many sites. Such information is helpful because it signals species that could
322	have their populations reinforced through management in order to strengthen
323	interactions at broader scales. In fact, common species may be good candidates for
324	population reinforcement rather than trophic rewilding <i>de facto</i> [48] when the focus is
325	to restore ecological services [5]. Those species are unlikely to be endangered and are
326	more likely to have sufficient captive stocks or source areas for translocation, which
327	makes them suitable candidates for such initiatives.
328	Although our framework provides a useful tool to identify rewilding candidates
329	within a region or a given site, the choice on whether a given species is indeed a good

candidate for trophic rewilding is not based solely on interaction patterns. Conservation 330 planning must encompass a careful assessment to determine if the amount of remaining 331 habitat is sufficient and if the fragments are suitable for the species to establish a viable 332 333 population, especially in severely fragmented landscapes [49]. Habitat suitability can 334 also be included as an additional layer of information that weights interaction 335 probabilities, what could alter the outcomes of the models providing more realistic 336 estimates. Evidently, a prerequisite for rewilding is that the underlying threat to the population is under control. Usual threats, such as hunting and predation pressure by 337 invasive species, need to be addressed before the start of any reintroduction, otherwise 338 the population is not viable in the long term. Finally, reintroductions are only feasible 339 when there are sufficient captive stocks or wild populations available for translocation, 340

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which is not the case for some of the species that score high in our study, such as *Brachyteles* spp. The objective of the proposed approach is not to replace any steps of
rewilding planning, but to offer one additional tool for conservation. Incorporating
predictions on ecological interactions into trophic rewilding planning is crucial and our
framework should be seen as a first step in ranking the ideal candidates for rewilding in
a given region. This list should then be filtered to species that fulfil the basic
requirements of any conservation translocation.

348 The lack of source populations was an issue faced by attempts of reintroducing of the red-billed curassow (Crax blumenbachii) in the Atlantic Forest, which was made 349 350 possible by efforts of captive breeding [50]. The most successful and widely known reintroduction program in the Atlantic Forest, which restored the endemic golden lion 351 tamarin (Leontopithecus rosalia) population from around 100 to more than 1600 352 individuals, only succeeded because it was built on a partnership between researchers, 353 conservation initiatives and zoos around the world. They committed to protect the 354 remaining habitat while reintroducing individuals [51]. Proportional efforts in terms of 355 resources are not always required when the focal species is a relatively common one, as 356 357 in the case of rewilding projects that aim to restore ecological processes and not to protect a given endangered species [52]. 358

By identifying the areas with the highest credit, the framework presented here may help to determine areas where trophic rewilding efforts should be targeted in order to maximize functional gains. Despite taking a number of simplifying steps, such as generating a naïve historical benchmark, we show in this first attempt that the proposed approach is a robust framework that helps producing testable predictions about the community responses to species reintroductions, besides aiding decision-making in conservation planning. We worked at a broad spatial scale encompassing the Atlantic

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366	Forest as a whole, but the same approach can also be useful when analysing the local
367	interaction credit, aiding in selecting candidate species and then evaluating success.
368	Although this framework has never been used before to prioritize areas and species for
369	rewilding, the concept of interaction credit has already been effectively used to predict
370	and evaluate reintroduction success. In a protected area within the Serra do Mar sub-
371	region, reintroduction of agoutis (Dasyprocta leporina) and howler monkeys (Alouatta
372	guariba), both important seed dispersers which had been locally extirpated, were
373	assessed in terms of interaction restoration, and a large proportion of the interaction
374	credit identified for the area was cashed in the first years after the release [12,53].
375	We focused here on seed dispersal interactions in a tropical biodiversity hotspot,
376	but the proposed framework can be applied to different biomes or types of interaction.
377	Even when considering other types of interactions, prioritizing the reintroduction of
378	species that are able to cash a high credit of ecological interactions may be important to
379	restore ecosystem functioning. Providing conservation practitioners with sufficient
380	information on how to prioritize species and areas in terms of ecological interaction
381	restoration is crucial for reversing the consequences of defaunation.
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383	Acknowledgements

We thank Elisabeth S. Bakker and Jens-Christian Svenning for inviting us to contribute to this Theme Issue. We thank the two anonymous reviewers for their comments that improved our manuscript. We also thank Laurence Culot and Milton Ribeiro for sharing data and the REFAUNA team for previous refaunation discussions.

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389 Data accessibility

2 3	390	All data used in this manuscript is published as data papers from the ATLANTIC series
4 5	391	data papers: https://github.com/LEEClab/Atlantic_series; and the Neotropical Tree
6 7 8	392	Communities database TreeCo version 2.0:
8 9 10	393	http://labtrop.ib.usp.br/doku.php?id=projetos:treeco:start. The packages and codes for
11 12	394	joint distribution models are available at
13 14	395	https://www.helsinki.fi/en/researchgroups/statistical-ecology/hmsc
15 16 17	396	
18 19	397	Author's contributions
20 21	398	MG, LG, ASP and FASF conceived the idea; EM, LG, MMP and MG designed the
22 23	399	study; RAFL, AAO and PIP compiled the tree data; EM conducted analyses; EM, LG
24 25 26	400	and MMP wrote the first draft of the manuscript, and all authors contributed
26 27 28	401	substantially to the final manuscript.
29 30	402	
31 32	403	Competing interests
33 34	404	We have no competing interests.
35 36	405	
37 38	406	substantially to the final manuscript. Competing interests We have no competing interests. Funding This work was partly supported by the Research Council of Norway through its Centres
39 40 41	407	This work was partly supported by the Research Council of Norway through its Centres
42 43	408	of Excellence Funding Scheme (223257) to EM and OO. LG and MMP are funded by
44 45	409	Coordenação de Aperfeiçoamento Pessoal de Nível Superior (CAPES). LG is funded by
46 47	410	Fundação de Amparo à Pesquisa do Estado do Rio de Janeiro (FAPERJ). FASF and
48 49	411	ASP receive personal grants by Conselho Nacional de Pesquisa e Desenvolvimento
50 51	412	Científico e Tecnológico (CNPq). MG receives a CNPq fellowship and is supported by
52 53 54	413	FAPESP grant 2014/01986-0, São Paulo Research Foundation (FAPESP). FASF, ASP
55 56	414	and LG receive support from Fundação Grupo Boticário de Proteção à Natureza
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http://labtrop.ib.usp.br/doku.php?id=projetos:treeco:start. The packages and codes for oint distribution models are available at https://www.helsinki.fi/en/researchgroups/statistical-ecology/hmsc Author's contributions AG, LG, ASP and FASF conceived the idea; EM, LG, MMP and MG designed the tudy; RAFL, AAO and PIP compiled the tree data; EM conducted analyses; EM, LG nd MMP wrote the first draft of the manuscript, and all authors contributed ril of ubstantially to the final manuscript. **Competing interests** We have no competing interests. Funding This work was partly supported by the Research Council of Norway through its Centres of Excellence Funding Scheme (223257) to EM and OO. LG and MMP are funded by Coordenação de Aperfeiçoamento Pessoal de Nível Superior (CAPES). LG is funded by Fundação de Amparo à Pesquisa do Estado do Rio de Janeiro (FAPERJ). FASF and ASP receive personal grants by Conselho Nacional de Pesquisa e Desenvolvimento Científico e Tecnológico (CNPq). MG receives a CNPq fellowship and is supported by FAPESP grant 2014/01986-0, São Paulo Research Foundation (FAPESP). FASF, ASP nd LG receive support from Fundação Grupo Boticário de Proteção à Natureza http://mc.manuscriptcentral.com/issue-ptrsb

415	(0010/2014), CNPq (grant numbers: 487092/2012-4 and 308356/2014-4) and FAPERJ					
416	(grant number: E-26/010/001645/2014). RAFL was funded by grant 2013/08722-5, Sa					
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576	Figur	e and table captions
577	Figur	e 1. Spatial credit of ecological interactions to be cashed through rewilding across
578	predic	ction sites in the Atlantic Forest. We calculated the interaction credit as the
579	differe	ence between a historical benchmark of species occurrences and their current
580	predic	eted occurrences. The number of interactions expected to be restored by rewilding
581	is dep	icted by a colour gradient. The warmest colours represent credit hotspots, whereas
582	the co	ldest colours represent sites where there are fewer seed dispersal interactions to be
583	restor	ed through reintroductions.
584	Table	• 1. Interaction credit of the 12 highest scoring frugivore species across the
585	Atlant	tic Forest. We measured the interaction credit contribution as the mean expected
586	numb	er of interactions across prediction sites where the frugivore is predicted to occur
587	accord	ding to the benchmark scenario. The total credit is the sum of credit contribution
588	for all	sites. We ranked species according to the mean contribution and present the
589	minin	num and maximum of local ranks across all prediction sites and the standard

	590	deviation of the ranks.	The last column represents the IUCN	threat categories each
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- 591 species is assigned to: LC = Least concern, NT = Near threatened, VU = vulnerable, EN
- 592 = Endangered, CR = Critically endangered [36].

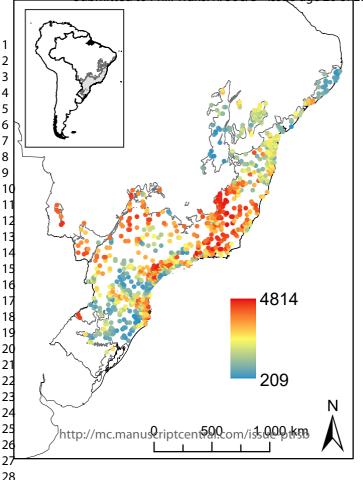
593 Figure 2. Distribution of the frugivore contributions to the credit of seed-dispersal

- 594 interactions across prediction sites in the Atlantic Forest. The large histogram in the
- 595 centre shows the distribution of the average credit contribution of each species
- 596 considering all sites. Smaller histograms show the distributions of credit contribution

597 for the 12 highest scoring species across prediction sites.

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Brachyteles arachnoides Penelope superciliaris Aburria jacutinga Carpornis cucullata Selenidera maculirostris Lipaugus lanioides Pteroglassus bailloni Tangara sayaca	47.99 34.04 30.45 28.68 28.33 27.76	11132.69 31048.42 24424.16 6652.67 24282.49	1-160 1-105 1-101 2-160 1-133	62.66 6.85 28.93 59.91 28.97	EN LC EN NT
Aburria jacutinga Carpornis cucullata Selenidera maculirostris Lipaugus lanioides Pteroglassus bailloni	30.45 28.68 28.33 27.76	24424.16 6652.67 24282.49	1-101 2-160	28.93 59.91	EN NT
Carpornis cucullata Selenidera maculirostris Lipaugus Ianioides Pteroglassus bailloni	28.68 28.33 27.76	6652.67 24282.49	2-160	59.91	NT
Selenidera maculirostris Lipaugus lanioides Pteroglassus bailloni	28.33 27.76	24282.49			
Lipaugus Ianioides Pteroglassus bailloni	27.76		1-133	28 97	1.0
Pteroglassus bailloni		47004.00		20.07	LC
-	00 55	17324.89	3-131	50.57	NT
Tangara sayaca	26.55	21907.07	2-116	25.5	NT
J	26.26	23946.78	1-120	10.78	LC
Brachyteles hypoxanthus	25.95	8354.46	3-155	59.99	CR
Tangara cyanoptera	25.33	10434.38	3-160	65.47	NT
Elaenia flavogaster	24.89	22702.16	1-106	14.19	LC
Ramphastos dicolorus	24.77	16768.37	3-138	47.69	LC
	Ramphastos dicolorus	Ramphastos dicolorus 24.77	Ramphastos dicolorus 24.77 16768.37	-	Ramphastos dicolorus 24.77 16768.37 3-138 47.69

