1	Short title: Patterns of taxonomic, functional and phylogenetic $\beta$ -diversity variation of
2	Neotropical small mammals
3	
4	Exploring patterns of taxonomic, functional and phylogenetic β-diversity variation
5	of Neotropical small mammals in a highly fragmented landscape
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18	Competing interests: The authors have declared that no competing interests exist.
19	

#### Abstract 20

Diversity can be partitioned in several components and dimensions that are affected in 21 22 different ways by habitat loss and fragmentation. However, these partitions and dimensions are rarely investigated on human-modified landscapes. In this study, we 23 investigated different partitions (Hill numbers) and dimensions (taxonomic [TBD], 24 functional [FβD] and phylogenetic [PβD]) of small mammal β-diversity in a fragmented 25 landscape of central Brazil using a multi-scale approach. TBD was estimated 26 27 considering rare, common and abundant species. Tolerance to disturbed habitat, assessed via the traits "habitat use", "tail length" and "use of vertical strata", and trophic 28 guild, defined by the "diet", were used to estimate FBD. PBD was based on phylogenetic 29 relatedness of the sampled species. The association between different partitions and 30 dimensions of β-diversity with habitat and landscape attributes were investigated using 31 32 Mantel tests. We found a significant positive effect of geographical distance on all partitions and dimensions of  $\beta$ -diversity. NDVI was the second most important variable 33 34 affecting abundance based T $\beta$ D, and all phylogenetic and functional  $\beta$ -diversity 35 dimensions. Habitat characteristics, such as fallen logs and canopy cover were positively associated with all β-diversity dimensions. Our findings support the 36 hypothesis that even in a highly modified landscape, small mammal's β-diversity is 37 38 determined by different environmental factors and spatial disposition of forest patches. However, the relatively higher importance of space appears to be related to dispersal 39 limitation of this group. 40

41

**Keywords:** abundance; geographical distance; NDVI; vegetation structure

# 42 Introduction

Biological diversity is a topic of great interest for biologists and can have different 43 connotations [1,2]. There are several ways to describe biological diversity, and among 44 them the definition of three main components: alpha diversity ( $\alpha$ ) - diversity on a local 45 scale, beta diversity  $(\beta)$  - the variation of species between locations, and gamma 46 diversity  $(\gamma)$  - diversity on a regional scale [3] have been largely employed. 47 Nevertheless, the traditional definition of  $\beta$ -diversity is dependent on  $\alpha$ -diversity [3,4], 48 49 leading to spurious results when researchers compare β-values of regions with different  $\alpha$ -diversities [5]. Therefore, to avoid misinterpretation,  $\alpha$  and  $\beta$  components should be 50 51 ideally transformed into their number equivalents [6,7]; for example  ${}^{q}D_{\beta}$  for  $\beta$ -diversity, where *q*-number could determine a diversity measure's sensitivity to rare or common 52 53 species [5,8]. In this approach, species richness, Hill-Shannon diversity and Hill-54 Simpson diversity [6] are the three forms of Hill diversity most commonly used in ecological studies [8] and are generally known as the "Hill numbers". 55 In a community dynamic viewpoint, β-diversity increases (heterogenization) 56 when common species do not co-occur from some or all sites, or when new species 57 58 arrive at some sites; and  $\beta$ -diversity decrease (homogenization) when rare, non-common species become extinct, or when formerly rare or absent species become widespread [9]. 59 60 Variation on the trajectory of  $\beta$ -diversity can be caused by different effects of human 61 disturbance [9–11]. Between these human disturbances, habitat loss has a consistently negative effect on biodiversity, while habitat fragmentation has been associated with the 62 concept of habitat spatial heterogeneity, generally considered to have a positive 63 influence on population and community-level ecological response [12,13]. In this 64

scenario of habitat fragmentation,  $\beta$ -diversity should increase due mainly the effects of environmental variation among patches.

Nonetheless, the accrued evidence showed that the patterns of  $\beta$ -diversity in 67 68 fragmented landscapes can result in homogenization or differentiation depending on the 69 landscape heterogeneity and the spatial scale of analysis [14–17], highlighting the importance of isolation by distance on determining patterns of β-diversity variation 70 71 [18]. Functional and phylogenetic  $\beta$ -diversity have been relatively less investigated on 72 fragmented landscapes when compared to traditional taxonomic metrics [19], but 73 despite less used, some studies show that a decrease in such dimensions of  $\beta$ -diversity 74 was associated with land use intensification [20,21]. On the other hand, in a gradient of 75 habitat complexity, the increase of all dimensions of  $\beta$ -diversity can be driven by  $\beta$ replacement (T $\beta$ D and F $\beta$ D) or  $\beta$ -richness (P $\beta$ D) [22]. So, ecological drives of 76 biodiversity can differ among different biotas, scales and diversity facets [23]. Thus, 77 decomposition of biodiversity into different dimensions may allow the identification of 78 79 the main aspects of communities that are affected by forest conversion and 80 fragmentation [24]. Therefore, in the abovementioned examples, habitat loss and fragmentation can lead to contrasting effects on  $\beta$ -diversity depending on the scale and 81 82 dimension investigated.

Previous studies in the Neotropical region have shown that small mammals are good model organisms for testing the influence of landscape heterogeneity and habitat complexity on community ecology [25–30]. However, studies conducted in a multiscale approach pointed contrasting scale-effects on small mammal  $\beta$ -diversity. For instance, in an Atlantic Forest fragmented landscape, small mammal  $\beta$ -diversity increased in small and isolated fragments [25], whereas, in Amazonian land-bridge islands, small mammal  $\beta$ -diversity was more strongly related to environmental variation

(habitat quality) among sites than to spatial distance, patch scale and landscape scale 90 91 [31]; further, in this ecoregion, all dimension of  $\beta$ -replacement decreased, while  $\beta$ richness increased with forest area [32]. In the Brazilian Savanna, spatial configuration 92 93 of the landscape and the extent, and quality of habitat strongly influence the rate of species turnover [30]. So, there is no consensus of the effects of landscape variation on 94 95 structuring the Neotropical small mammals  $\beta$ -diversity, clearly demonstrating the need 96 of further studies exploring different community dimensions to better understand the effects of habitat fragmentation on this important group of vertebrates. 97

98 In this study, we evaluated the response of species richness, Shannon's entropy 99 and Simpson's dominance on TBD (multiplicative partitioning of Hill numbers) [5] to 100 understand the relative importance of rare and dominant species on determining patterns 101 of β-diversity in a highly fragmented landscape. We also investigated which are the main predictors of habitat (local) and landscape (regional) scales affecting the partitions 102 of taxonomic, functional and phylogenetic  $\beta$ -diversity. Dispersal ability and habitat 103 104 selectivity have been described as the main processes related to change on species composition [28], functional structure [29] and species turnover [30] of small mammal 105 106 community in highly fragmented landscapes. Considering the relative low dispersal 107 ability of small mammals [33-35], and the effect of habitat and landscape scale on 108 structuring small mammal composition in Neotropical region [26,28,30,31], we would expect a positive effect of space, habitat and landscape quality on influencing all 109 partitions and dimensions of  $\beta$ -diversity. 110

111

# **Materials and Methods**

## 113 Study area

- 114 The study was conducted in southern Goiás State, central Brazil (18°25' 18°43' S,
- $115 \quad 50^{\circ}48' 50^{\circ}22'$  [28]), a highly fragmented landscape with about 13% of forest cover
- 116 (https://mapbiomas.org/), dominated by semi-deciduous forest connected to riparian
- 117 forest in a transitional region between Atlantic Forest and Cerrado ecoregion [36]. The
- 118 climate is Tropical semi-humid Aw (Köppen) with markedly dry season (April to
- 119 September) and wet season (October to March), and mean annual temperature and mean
- annual rainfall around 23 °C and 1600-1900 mm, respectively [37].

121

## 122 Sampling design

The sampling was carried out between January and December 2015. We captured small 123 124 mammals, and quantified the vegetation structure and food resources in 24 trapping grids, distant 0.5 to 1 km within the same patch. Each trapping grid was composed of 20 125 trap stations, representing an area of 45 x 60 m (see, [28] for details about trapping 126 grids). For landscape characterization, we estimated landscape metrics in 13 buffers of 1 127 km radius from the edge of sampled forest patches. The average forest patch size was 128 84.1 ha, with the smallest patch with 39.5 ha and the largest one with 142 ha. The 129 average Euclidean distance among them was 22,588 m, being the closest distant 389 m 130 and the farthest 51,978 m. 131

#### 133 Small mammal surveys

134	We captured small mammals using live traps (wire-cage traps and Sherman traps) and
135	pitfall traps. We set Sherman and wire-cage traps in 16 trap-stations arranged in four
136	linear transects, with 15-m intervals between the nearest trap stations. Each trap-station
137	was composed by one wire-cage and one Sherman trap disposed on ground and
138	understory (1.5 to 2 m height), alternatively. Pitfall traps was arranged in a
139	perpendicular transect distant 15 m of each live trapping-grid, composed by four
140	buckets (30 liters) connected by a fence of 0.8 m height (see, [28] for methodological
141	detail). We made the captures under the collection license SISBIO nº 46985-1 and in
142	accordance with guidelines provided by the American Society of Mammalogists [38].

143

## 144 Habitat and landscape characterization

We used 23 variables to describe habitat and landscape characterization (S1 Data). For 145 146 habitat characterization, we measured six variables related to vegetation structure (numbers of trees, shrubs, lianas, fallen logs, canopy cover and litter cover) and two 147 148 variables that depict food resource availability (arthropods and fruits and seeds 149 composition) in 10 selected trap stations of each trapping grid. We reduced the dimensionality of food resource availability using a Principal Coordinates Analysis 150 (PCoA) separately for the matrix of arthropods and fruits-seeds resources. We 151 152 associated the both matrices with the "horn" dissimilarity index, the one that presented higher variance recovery in relation to the original distances. As for the patch scale, we 153 154 used four variables (i.e., forest area, core area, perimeter and NDVI), whereas for landscape scale, we used nine variables (i.e., mean isolation, water distance, shannon 155 diversity, shape, mean perimeter area, core area, index core area, total edge and 156

157 connectivity). The attributes related to patch and landscape scales were obtained via158 satellite images (methodological details can be found in [28]).

159

## 160 Functional traits and phylogeny building

We selected four functional traits associated with tolerance to habitat disturbance: 1) habitat use [nominal trait: generalist or forest specialist], 2) tail length [quantitative trait in mm: arboreal species tend to be long tailed], 3) understory use [quantitative trait in %: based on the percentage of capture in the understory]; and trophic guild: 4) diet [multi-choice: insectivore, frugivore, granivore and omnivore, according to Annotated Checklist of Brazilian Mammals [41] (see, [29] for functional traits interpretation details). Traits were measured in the field from captured specimens or obtained from the

168 literature [26,39–42].

We derived a phylogeny for our data set using the VertLife.org phylogeny subset on-line tool. We used as backbone the "Mammal's birth-death node-dated completed trees (all 5911 species, set of 10k trees)" tree. Despite the criticisms regarding the use of synthesis-based phylogenies in evolutionary community studies, mostly because the relative low resolution and higher number of polytomies [43,44] demonstrated that this method of generating phylogenetic trees is sufficiently robust for community phylogenetic analysis.

176

# 177 Data analysis

We initially tested for multicollinearity among all the 23 predictor variables describedabove using the Variance Inflation Factor (VIF) approach. For this, we used a stepwise

procedure, where all data set was tested for collinearity, and when a variable showed 180 181 VIF value above the threshold of 10, it was excluded and the procedure repeated with remaining variables until no variable was excluded. After such procedure, a total of 15 182 variables remained to be used as predictors in our models (S1 Table). 183 184 We used various distance-based metrics at the taxonomic, functional and phylogenetic dimensions to test for variation of β-diversity along the predictor variables 185 186 selected above. For the taxonomic dimension, we partition the diversity of the metacommunity into  $\beta$ -components weighting different orders of diversity: species 187

richness [q = 0], Shannon's diversity [q = 1] and Simpson's dominance [q = 2] (Hill

numbers, [5-7]). For each Hill number, we did such partition for each pair of grids; thus,

190 we estimate the  $\beta$ -diversity variation among all possible pairs of grids to generate a

191 distance-based triangular matrix of  $\beta$ -diversity.

As for the functional dimension, we associated each pair of species by the eight 192 193 continuous traits using the "Gower" distance. We then used this distance-based 194 triangular matrix to estimate the Rao's quadratic entropy [45], a measure of diversity in ecological communities accounting for species differences (functional or phylogenetic). 195 We estimate two metrics of FBD: Dkl, which is the pairwise functional distance, and H, 196 197 which is the Dkl standardized to account for within-community diversity. Both metrics return distance-based triangular matrix of FBD among pair of grids. For the 198 phylogenetic dimension, we used the phylogenetic relationship of the captured species 199 200 of small mammals to calculated the mean phylogenetic distance among all pairwise 201 combinations of species co-occurring in a sample (MPD), which is a basal measure of 202 the phylogenetic relatedness, and phylogenetic distance between each species and its nearest neighbor on the phylogenetic tree (MNTD), which is can be interpreted as a 203 terminal metric of the phylogenetic relatedness of co-occurring species [46,47]. 204

To test the association between each of the explanatory variables with the 205 different dimensions of  $\beta$ -diversity, we used Mantel tests with 9999 permutations. For 206 this, we used each of the  $\beta$ -diversity distance based on the triangular matrix as a 207 208 response matrix, and a Euclidean distance-based on a matrix associating each sample unit by each predictor as a predictor matrix. We have also used a triangular matrix based 209 on the Euclidean distance among grids to test for spatial autocorrelation. Moreover, 210 211 given that our dataset is composed by 13 forest patches, being 12 of them with two grids, for testing the effect of the landscape metrics, we sampled one trapping grid per 212 forest patch, thus reducing the number of sampling units to 13 to avoid pseudo-213 214 replication at a landscape scale. In the end, we did a total of 16 mantel tests for each  $\beta$ diversity dimension described above. Finally, we tested for the spatial autocorrelation of 215 216 each of the 15 predictor variables, and whenever we found an association between a  $\beta$ -217 diversity metric and any metric spatial autocorrelated, we did partial mantel tests in order to account for any potential confounding effect between explanatory variables. 218 219 We did all our analyses and graphics in R version 4.0.5 [48]. For calculating 220 the VIF we used the package "*usdm*" [49]. To calculate the pairwise  $\beta$ -diversity at the taxonomic level, we used the function "DivPart" of the package "entropart" [50]. To 221 associate our functional matrix based on "Gower" distance, we used the function 222 "vegdist", and to perform the Mantel tests, we used the "mantel" function, both from the 223 package "*vegan*" [51]. To estimate the Rao's quadratic entropy, we used the function 224 "raoD", and to calculate "MPD" and "MNTD" we used the functions "mpd" and "mntd", 225 respectively, all from the "picante" package [52]. 226

227

# 228 **Results**

#### 229 Small-mammal diversity

- 230 With an effort of 12,096 trap-night, we had 624 captures, resulting in a trap success of
- 5.2%. We captured 408 individuals (mean  $\pm$  standard deviation (SD),  $16.67 \pm 8.76$
- individuals per trapping grid), belonging to 12 small mammal species  $(4.58 \pm 1.50)$
- species per trapping grid). The most common species were the arboreal marsupial
- 234 Gracilinanus agilis (124 individuals, 20 trapping grids); the arboreal rodents
- Rhipidomys macrurus (55, 14), Oecomys bicolor (52, 16) and Oecomys catherinae (27,
- 6); the scansorial marsupials *Didelphis albiventris* (49, 20) and *Marmosa murina* (28,
- 8); and the terrestrial rodent *Calomys expulsus* (55, 13), comprising 96.1% of all
- individuals captured. The rarest species represented by less than 20 individuals were the
- terrestrial rodent *Calomys tener* (6 individuals, 4 trapping grid); the scansorial
- 240 marsupial *Cryptonanus chacoensis* (5, 4); the scansorial rodent *Oligoryzomys*
- 241 *mattogrossae* (2, 2); the terrestrial marsupial *Monodelphis kunsi* (2, 2); and the arboreal
- 242 marsupial *Caluromys philander* (1, 1).

243

## 244 Explanatory variables variation

For habitat characterization – vegetation structure and food resource availability, we found the following amplitude variation: "no. of shrubs" (min = 62, max = 1027), "no. of lianas" (min = 86, max = 418), "no. of fallen logs" (min = 3, max = 58), "canopy cover" (min = 41%, max = 98%) and "litter cover" (min = 85%, max = 99%). We considered the first two PCoA axes obtained from "arthropods" resource matrix

250	(variance recovery, $r^2 = 0.85$ ), and the first two PCoA axes from the "fruit-seed"
251	resource matrix ( $r^2 = 0.87$ ) to represent "food resource availability" (S2 Table).
252	Landscape characterization was based on patch and landscape metrics, according to
253	following amplitude variation parameters: "forest area of focal patch" (min = 39.6, max
254	= 142.3 ha), "perimeter length of focal patch" (min = 2749, max = 21136 m), "NDVI"
255	(min = 0.2800, max = 0.3370), "patch isolation" (min = 99, max = 4102 m), "water
256	distance" (min = 171.6, max = 1258.4 m) and "total edge of landscape" (min = 6410,
257	max = 60766 m). Average values for all variables patch and landscape variables can be
258	found elsewhere (S2 Table). Within the 15 predictor variables, we have only found
259	evidence that "forest area" was spatially autocorrelated (Mantel test, $r = 0.48$ , $p <$
260	0.001). All other variables showed no signal of spatial structure (S3 Table).

261

#### Functional and phylogenetic variation 262

263 The small mammal community represented a range of different ecological traits, seven

species were classified as habitat generalist (e.g., marsupials [C. chacoensis, D. 264

albiventris, G. agilis and M. kunsi] and rodents [C. expulsus, C. tener, O. 265

*mattogrossae*]), and five species were forest specialist (marsupials [C. philander and M. 266

murina] and rodents [O. bicolor, O. catherinae and R. macrurus]). In relation to tail 267

268 length, the shorter and longer tails were found in the terrestrial M. kunsi (36.1 mm) and

- scansorial D. albiventris (297.8 mm) marsupials, respectively. Thus, tail length of 269
- marsupials ranged from 36.1 to 297.8 (mean  $\pm$  SD, 156  $\pm$  87.8 mm), while the tail 270
- 271 length of rodents ranged from 58.6 to 148.2 (96.8  $\pm$  34.3 mm). The use of understory
- ranged from 0 (M. kunsi, C. expulsus, C. tener and O. mattogrossae) to 100% (C. 272
- *philander*), with little variation within each small mammal group (marsupials: min = 0, 273

274	max = 100, 52.3 $\pm$ 41.7; rodents: min = 0, max = 93.1, 42.2 $\pm$ 47 frequency). We found
275	no association between tail length and understory use (r = 0.44, $df = 10$ , p = 0.15),
276	showing that these traits are complementary to described use of vertical stratum. For
277	trophic guild, the small mammals captured species were assigned to a diet category
278	representing a combination of feeding guilds, such as: frugivore-omnivore (C. philander
279	and D. albiventris) and insectivore-omnivore (C. chacoensis, G. agilis, M. murina and
280	M. kunsi); while rodents were classified as: frugivore-granivore (C. expulsus, C. tener
281	and O. mattogrossae) and frugivore-seed predator (O. bicolor, O. catherinae and R.
282	macrurus). Categories were based on Paglia et al. [41].
283	The marsupial species captured in our study belong to the family Didelphidae,
283 284	The marsupial species captured in our study belong to the family Didelphidae, and are distributed into two subfamilies: Caluromyinae (i.e., <i>C. philander</i> ) and
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284 285	and are distributed into two subfamilies: Caluromyinae (i.e., <i>C. philander</i> ) and Didelphinae that can be sub-divided in three tribes: Marmosini ( <i>M. murina</i> and <i>M</i> .
284 285 286	and are distributed into two subfamilies: Caluromyinae (i.e., <i>C. philander</i> ) and Didelphinae that can be sub-divided in three tribes: Marmosini ( <i>M. murina</i> and <i>M. kunsi</i> ), Didelphini ( <i>D. albiventris</i> ) and Thylamyini ( <i>C. chacoensis</i> and <i>G. agilis</i> ). The
284 285 286 287	and are distributed into two subfamilies: Caluromyinae (i.e., <i>C. philander</i> ) and Didelphinae that can be sub-divided in three tribes: Marmosini ( <i>M. murina</i> and <i>M. kunsi</i> ), Didelphini ( <i>D. albiventris</i> ) and Thylamyini ( <i>C. chacoensis</i> and <i>G. agilis</i> ). The six rodent species comprised the suborder Myomorpha, family Cricetidae, subfamily
284 285 286 287 288	and are distributed into two subfamilies: Caluromyinae (i.e., <i>C. philander</i> ) and Didelphinae that can be sub-divided in three tribes: Marmosini ( <i>M. murina</i> and <i>M. kunsi</i> ), Didelphini ( <i>D. albiventris</i> ) and Thylamyini ( <i>C. chacoensis</i> and <i>G. agilis</i> ). The six rodent species comprised the suborder Myomorpha, family Cricetidae, subfamily Sigmodontinae, belonging to three tribes: Oryzomyini ( <i>O. bicolor, O. catherinae</i> and <i>O</i> .

292

# 293 **Taxonomic β-diversity**

The comparison of T $\beta$ D variation with our predictor variables showed consistent results when weighting the species richness (q = 0) and Shannon entropy (q = 1). We found evidence of geographical distance (q = 0: r = 0.38, p = 0.008 [13 sample units]; q = 0: r

297	= 0.32, p = 0.001 [24 sample units]; q = 1: r = 0.27, p = 0.047 [13 sample units]; q = 1: r
298	= 0.34, p = 0.001 [24 sample units]; Fig 2a and b) on the patterns of $\beta$ -diversity (Table
299	1). We have not found any other association with else predictor variables for these $\beta$ -
300	diversity metrics. For the $\beta$ -diversity weighting the dominant species (q = 2), we found
301	that geographical distance (r = 0.27, p = 0.042 [13 sample units]; r = 0.26, p = 0.004 [24
302	sample units]; Table 1, Fig 2c), "no. of fallen logs" (r = 0.22, p = 0.014; Table 1; Fig
303	2d) and "canopy cover" ( $r = 0.19$ , $p = 0.038$ ; Table 1, Fig 2e) were important predictors
304	on determining the patterns of $\beta$ -diversity turnover. Despite occurring all over the
305	gradient, higher fallen logs frequency seems to determine the higher incidence of C.
306	tener, M. murina and R. macrurus, and lower fallen logs frequency appears to
307	associated to G. agilis and C. philander (S1 Fig). As for the canopy cover, most of the
308	species were relatively abundant along the entire gradient, but some species like M.
309	kunsi, O. mattogrossae and C. philander were exclusively found in grids with higher
310	canopy cover, and C. expulsus was more frequently capture in low canopy cover sites
311	(S2 Fig).

312

# **Functional and phylogenetic β-diversity**

The functional turnover represented by Dkl was positively associated with "no. of fallen logs", "canopy cover" and "NDVI" (r = 0.15, p = 0.030; r = 0.18, p = 0.004; r = 0.25, p = 0.018, respectively; Table 1, Fig 3a-c). In the other hand, the standardized H was related only to geographical distance (r = 0.36, p = 0.013; Table 1, Fig 3d). Higher values of Dkl and H were associated with species like *O. mattogrossae*, *D. albiventris*, *C. chacoensis* and *O. bicolor*; species with a wide range of variation in their traits

related to use of habitat, tail length, and diet, which have determined such high valuesof functional diversity variation (S3 Fig).

322	The P $\beta$ D represented by MPD showed a positive association with geographical
323	distance and "forest area" (r = 0.36, p = 0.006 [13 sample units]; r = 0.34, p = 0.001 [24
324	sample units]; $r = 0.42$ , $p = 0.017$ ; Table 1, Fig 4a and b). The effect of "forest area" is
325	consistently significant even when accounting for the variability captured by the
326	geographical disposition of the grids (Partial mantel, $r = 0.31$ , $p = 0.04$ ). The MNTD
327	showed significant associations with geographical distance and "canopy cover" ( $r =$
328	0.51, p = 0.007 [13 sample units]; r = 0.46, p = 0.001 [24 sample units]; r = 0.32, p =
329	0.030; Table 1, Fig 4c and d). Marsupials, Marmosini tribe (M. murina and M. kunsi)
330	and arboreal rodents, tribes Oryzomyini (O. bicolor) and Thomasomyini (R. macrurus)
331	were more frequent in habitats with high "canopy cover" values (S2 Fig).

332

# 333 Discussion

## 334 Overview

Our findings revealed that *i*) small mammal communities sampled in this highly fragmented landscape Neotropical region were dominated by commonly found marsupials and rodent's species with arboreal-scansorial habits. *ii*) In spite of being highly fragmented, our landscape presented a wide variation of habitat quality, mostly driven by vegetation structure, food resource and landscape attributes, which harbor functionally and phylogenetically diverse small mammal communities in central Brazil. *iii*) Geographical distance was the main variable correlated to all partitions and all

342	dimensions of $\beta$ -diversity. Habitat quality gradient was an important predictor for T $\beta$ D
343	(partition of dominant species), and F $\beta$ D and P $\beta$ D. We discuss below such findings in
344	the light of dispersal limitation and environmental filters on driving the patterns of $\beta$ -
345	diversity in small mammals of Central Brazil.

346

## **347 Taxonomic β-diversity**

348 Geographical distance was positively correlated with all partitions of  $T\beta D$ ,

349 corroborating our initial hypothesis. Marsupials and small rodents have a low dispersal

ability [33,35,51,52] that limit the continuous replacement of small mammal species

351 between distant localities. Further, small mammal communities in the Cerrado domain

have been discussed elsewhere to be mainly driven by dispersal limitation and habitat

selectivity [28,30]. Geographical distance also has been the most important predictor of

small mammal dissimilarity in Atlantic Forest [52]. This finding can be explained by

the idea that fragmented landscapes are hyper-dynamically influenced by environmental

heterogeneity [12]. Small mammals are able to disperse between relatively closer

357 fragments (~ 485 m in average, according to [55]), but homing behavior can be much

higher for larger species like *Philander frenatus*, which crossed an area of 1050 m of

359 fragmented landscape in the Atlantic Forest [56]. In our study, although the landscape is

360 mainly composed by homogeneous semideciduous forest fragments, strong

361 heterogeneity at small scales can be found between forest patches (e.g., vegetation

362 structure). Such variability may reflect different levels of habitat quality, which

363 certainly influence the potential source pools and species establishment for each patch

364 with different degrees of isolation.

When weighting abundant species (q = 2 Hill number), T $\beta$ D was also positively 365 366 associated to habitat quality (e.g., fallen logs and canopy cover). We captured an expressive number of individuals in such highly fragmented landscape of central Brazil, 367 within these only C. philander, M. kunsi and O. mattogrossae were rarely recorded in 368 our study (1 to 2 individuals). On the other hand, 96% of the captured individuals -369 seven of the twelve captured species - occurred in 6 to 20 trapping grids. This finding 370 371 may be caused by the variation in vegetation structure between different sampling sites, which by consequence determine abundance patterns for the most common species. 372 373 Thus, the higher values of T $\beta$ D based on the abundance of species of small mammals 374 can be resulted from the higher variability in small mammal community structure along this habitat quality gradient. In the Cerrado domain, the abundance (not richness) of 375 376 small mammals is known to be correlated with different types of vegetation cover (e.g., 377 herbaceous, shrub, and tree density) [57], a variable that is commonly associated with habitat heterogeneity and complexity [58]. 378

379

## **Functional and phylogenetic β-Diversity**

In overall terms, the small mammals communities sampled in our study is relatively 381 species poor when compared to other communities of the same group in Central Brazil 382 383 [59], a pattern that seems to related to the elevated habitat loss in the region. Large bodied rodents that are specialists in forested habitats (e.g., tribe Oryzomyini [Hylaemys 384 megacephalus], Family Echimyidae, Eumysopinae [Proechimys longicaudatus and P. 385 386 *roberti*]) generally represent the first functional group and respective phylogenetic lineages to disappear in highly fragmented landscapes [25,60]. Such group is intimately 387 associated with continuous forests and landscapes with high percentual of vegetation 388

cover [25,27,39,60,61]. In the other hand, species of marsupials (e.g., G. agilis e C. 389 390 chacoensis) and rodents (e.g., C. expulsus e C. tener) that have a broad ecological niche 391 and are able to use both forested and open field environments, as present an 392 opportunistic feeding habitat [29,60], commonly represent the species that are capable to be successful even in such disturbed environments. Small sized rodent species that 393 394 use the understory (e.g., O. bicolor e R. macrurus) seems also to be weakly affected by 395 habitat fragmentation in the region, perhaps the smaller scale of niche requirements of 396 these species protected them of the negative effects of habitat fragmentation [29]. 397 However, despite high co-occurrence of such resistant species between forest fragments 398 in the region, variation in abundance patterns are common for these species, which may be related to different niche requirements and responses to habitat modification [29]. 399

Despite the relatively modest phylogenetic and functional variation of our 400 401 sampled small mammals' communities, we found a turnover for these community 402 dimensions in relation to geographical distance and habitat quality gradient (i.e., no. fallen logs and canopy cover). In the Amazonian habitat, FBD and PBD of bat 403 404 communities were highest between continuous forest and *campinarana*, two highly 405 contrasting environments, and these patterns in  $\beta$ -diversity were associated with 406 functional richness and lineage richness differences, respectively [22]. An study with 407 bird and ant assemblages, in Atlantic Forest and Pampas Grassland showed that land uses and biomes seems to promote assemblage differentiation in traits and lineages that 408 409 occurred in anthropogenic habitats, further in this landscape both animal groups were 410 similarly sensitive to changing in vegetation structure [63]. Therefore, habitat quality seems to be an important driver of functional and phylogenetic turnover in biological 411 communities in the Neotropics independent of the biological group or sampled region. 412 Nevertheless, understanding what are the most important traits and functional lineages 413

that readily respond to habitat disturbance, as what are the most resilient ones that
persist in disturbed fragments, are of utmost importance for providing information for
conservation initiatives.

Little is known about the dispersibility of small mammals among forest patches 417 in the Neotropics [55]. Despite reasonable to expect, a positive relation between body 418 419 size and dispersiveness is not always the rule for this group of mammals. For instance, the mid-bodied size nocturnal marsupial P. frenatus (400-600 g) has a relatively small 420 421 living area (2.8 ha). Thus, even being a larger species for a small-mammal, the apparent inability of this species to occupy different types of habitats (e.g. less forested patches 422 or open fields), may limit its dispersion to equivalent nearby fragments, letting this 423 424 species susceptible to local extinctions [64], which may similarly be the case for other 425 large bodied terrestrial small rodents that was expected to be reported in our study, but were not captured [e.g. Cerradomys scotti, C. maracajuensis, C. marinhus, C. 426 subflavus, P. longicaudatus, P. roberti]. The capacity of using the landscape matrix as 427 a secondary habitat, confers more flexibility to overcome the negative effects of habitat 428 429 fragmentation [65,66]. Therefore, the prevalence of mostly generalist's species that can occupy different portions of the landscape found here may explain the low FBD e PBD 430 in the sampling region. 431

432

# 433 Conclusion

To the best of our knowledge, this is the first study that shows the effect of geographical

distance, habitat and landscape variation on all partitions and dimensions of small

436 mammals'  $\beta$ -diversity. Considering that we have a depleted small mammal community

due to high habitat loss in central Brazil, we conclude that only abundance-weighted β-437 438 diversity values responded to the predictors of habitat quality. In this sense, a homogeneous environment results in poor communities, where abundance was more 439 important than richness, therefore, studies that seek to investigate the effect of habitat 440 loss and fragmentation on biodiversity need to consider species abundance and not just 441 species richness. Abundance was also a more important attribute than richness in anuran 442 443 assemblages in Pantanal, a region naturally disturbed by floods in South America, since most species co-occurred in all sampled ponds but varied in their abundance [67]. Thus, 444 we summarize that taxonomic, functional and phylogenetic replacement of the small 445 446 mammal communities in the fragmented landscape of central Brazil is determined by dispersal limitation and habitat selection, driven by species-specific responses in the 447 communities' arrangements. 448

449

# 450 Acknowledgments

We thank to the colleague of the Laboratory of Ecology and Biogeography of Mammalsfor field assistance.

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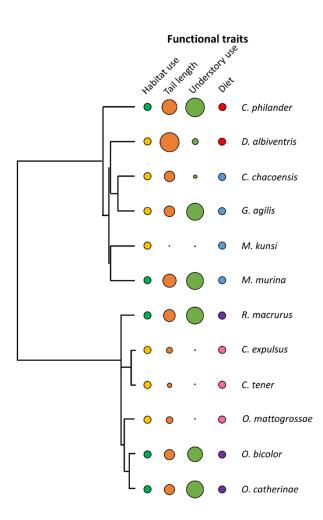
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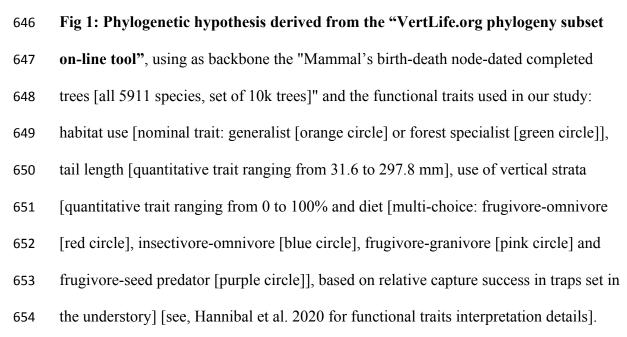
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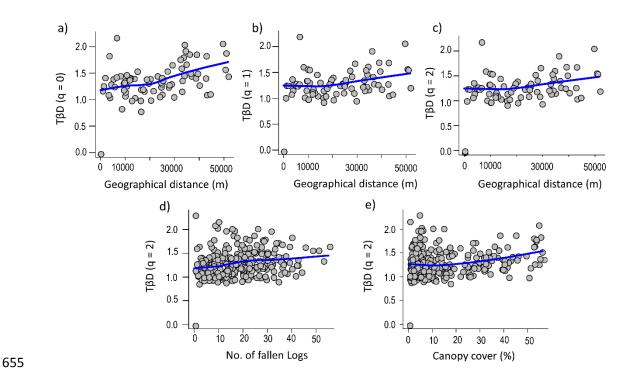
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#### 644 Figures.

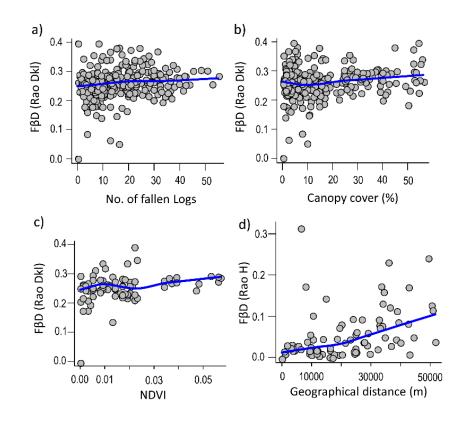






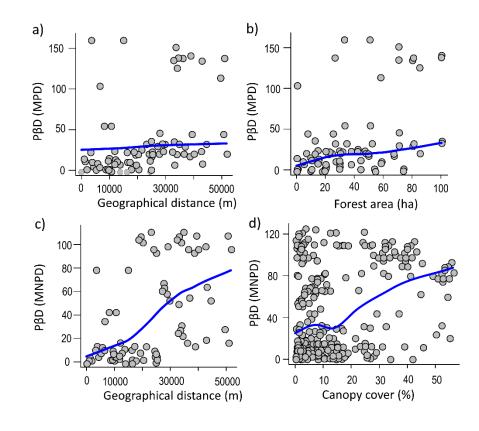
**Fig 2. Correlation between small mammal taxonomic β-diversity and geographical** 

- distance [a-c], no. of fallen logs [d] and canopy cover [e] in a highly fragmented
- **landscape in central Brazil.** The Hill numbers associated represent rare [q = 0],
- 659 common [q = 1] and abundant [q = 2] species, respectively. The tendency line is fitted
- using Locally Estimated Scatterplot Smoothing [LOESS].



662 Fig 3. Correlation between small mammal functional β-diversity and no. of fallen

- logs [a], canopy cover [b], NDVI [c] and geographical distance [e] in a highly
- 664 **fragmented landscape in central Brazil**. The tendency line is fitted using Locally
- Estimated Scatterplot Smoothing [LOESS].



**Fig 4. Correlation between small mammal phylogenetic β-diversity and** 

- 668 geographical distance [a and c], forest area [b] and canopy cover [d] in a highly
- 669 **fragmented landscape in central Brazil.** The tendency line is fitted using Locally
- 670 Estimated Scatterplot Smoothing [LOESS].

671 Table.

- 672 Table 1. Pearson correlation between geographical distance, habitat
- 673 characterization [vegetation structure and food resource availability] and
- 674 landscape characterization [patch and landscape metrics] with taxonomic,
- 675 functional and phylogenetic small mammal β-diversity. The significance of the
- 676 relation was tested via Mantel-tests.

	Taxonor	Taxonomic [Hill numbers]		<b>Functional Rao</b>		Phylogenetic	
	q = 0	q = 1	q = 2	Dkl	Н	MPD	MNTD
Geographical	0.38*	0.27*	0.27*	0.19	0.36*	0.36*	0.51*
distance [13							
sample units]							
Geographical	0.32**	0.34**	0.26**	0.19*	0.36**	0.34**	0.46**
distance [24							
sample units]							
Vegetation structure	e						
No. shrubs	-0.15	-0.08	-0.02	0.07	0.01	-0.06	-0.05
No. lianas	-0.06	-0.05	-0.02	-0.02	-0.12	0.00	-0.12
No. fallen logs	-0.08	0.13	0.22*	0.15*	0.05	-0.08	0.02
Canopy cover	0.01	0.15	0.19*	0.18**	0.15	-0.06	0.32*
Litter cover	-0.06	-0.02	-0.01	0.01	-0.03	0.06	0.17
Food resource avail	ability						
PCoA Arthro1	-0.05	-0.04	-0.03	-0.04	-0.02	0	0.12
PCoA Arthro2	-0.04	-0.09	-0.06	0.03	-0.08	-0.07	-0.13
PCoA Fruit1	-0.03	0.00	-0.02	-0.04	-0.05	-0.04	-0.07
Patch metrics							
Forest area	0.17	0.06	0.09	-0.14	0.23	0.42*	0.16
Perimeter	0.08	0.09	0.05	0.04	-0.13	-0.13	-0.11
NDVI	-0.06	0.13	0.13	0.25*	0.18	-0.00	0.39

Landscape metrics							
Isolation	-0.22	-0.25	-0.25	-0.16	-0.14	-0.15	-0.09
Water distance	-0.19	-0.29	-0.29	-0.13	-0.19	-0.17	-0.10
Total edge	0.05	0.01	0.01	0.07	-0.07	-0.10	-0.12
Connectivity	0.05	0.02	0.02	-0.05	-0.02	0.01	0.04

677 Significant p-value = \*<0.05, \*\*<0.01

# 679 Supporting Information

#### 680 S1 Table. VIFs of the remained variables after stepwise procedure, based on 23

#### 681 variables that characterized habitat and landscape scales.

Variables	VIF
No. shrubs	3.196
No. lianas	2.673
No. fallen logs	3.750
Canopy cover	2.594
Litter cover	2.741
PCoA Arthro1: Arthropods resources according to axis 1 of PCoA	2.464
PCoA Arthro2: Arthropods resources according to axis 2 of PCoA	3.561
PCoA Fruit-Seed1: Fruit-seed resources according to axis 1 of PCoA	2.620
Forest area of patch [ha]	3.184
Perimeter of focal patch [m]	8.916
NDVI	3.557
Isolation: mean isolation of the five nearest patches [m]	1.957
Water distance: distance to water course [m]	3.780
Total edge in the landscape [m]	5.172
Connectivity [connected versus isolated patch, considering riparian forest]	4.424

682

#### 684 S2 Table. Habitat, patch and landscape metrics that characterization fragmented

#### 685 landscape studied.

		Description	Range	Mean $\pm$ SD
	parameter			
No. shrubs	Vegetation	Number of shrubs in	62-1027	228.60±193.01
	structure	trapping-grid		
No. lianas	Vegetation	Number of lianas in	86-418	224.27±80.04
	structure	trapping-grid		
No. fallen	Vegetation	Number of fallen logs in	3-58	33.63±14.41
logs	structure	trapping-grid		
Canopy	Vegetation	Canopy cover in trapping-	41-98%	85.92±14.70
cover	structure	grid		
Litter cover	Vegetation	Litter cover in trapping-grid	85-99%	95.55±3.32
	structure			
PCoA	Food resource	PCoA axis 1 of arthropods	-	-
Arthrol		in trapping-grid		
PCoA	Food resource	PCoA axis 2 of arthropods	-	-
Arthro2		in trapping-grid		
PCoA Fruit-	Food resource	PCoA axis 1 of fruit and	-	-
Seed1		seed in trapping-grid		
PCoA Fruit-	Food resource	PCoA axis 2 of fruit and	-	-
Seed2		seed in trapping-grid		
Forest area	Patch metric	Total area of sampled patch	39.6-142.3	81.52±36.23
		[ha]		
Perimeter	Patch metric	Perimeter of sampled patch	2749-21136	6189.2±4733.7
		[m]		
NDVI	Patch metric	Normalized difference	0.28-0.33	$0.32 \pm 0.01$
		vegetation index		

Isolation	Landscape	Mean distance of the 4	99-4102	1603.1±1311.5
	metric	nearest patches [m]		
Water	Landscape	Distance of trapping grid to	171.6-1258.4	628.7±344.7
distance	metric	water course [m]		
Total edge	Landscape	Total edge measured in the	6410-60766	23679.9±15560.5
	metric	landscape [m]		

686

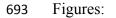
#### 688 S3 Table. Correlation between predictors variables and spatial distribution of the

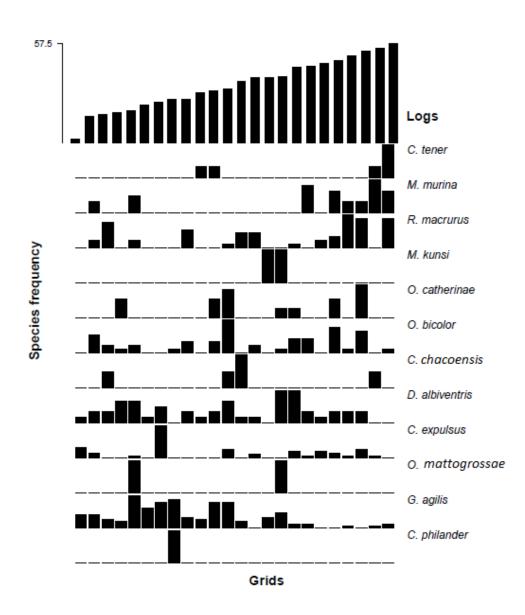
#### 689 sampled forest patches.

Group	Method	Statistic	N	P.value
No. shrubs	Pearson	0.0666	276	0.315
No. lianas	Pearson	-0.0088	276	0.509
No. fallen logs	Pearson	0.0312	276	0.317
Canopy cover	Pearson	0.0759	276	0.205
Litter cover	Pearson	0.0167	276	0.401
PCoA Arthro1	Pearson	0.0229	276	0.292
PCoA Arthro2	Pearson	-0.1020	276	0.837
PCoA Fruit-Seed1	Pearson	-0.0567	276	0.752
Forest area	Pearson	0.4495	78	0.007*
Perimeter	Pearson	-0.1249	78	0.745
NDVI	Pearson	-0.0000	78	0.458
Isolation	Pearson	-0.0210	78	0.502
Water distance	Pearson	-0.0370	78	0.572
Total edge	Pearson	-0.0896	78	0.721
Connectivity	Pearson	-0.0836	78	0.831

690

#### 692 Supporting Information:



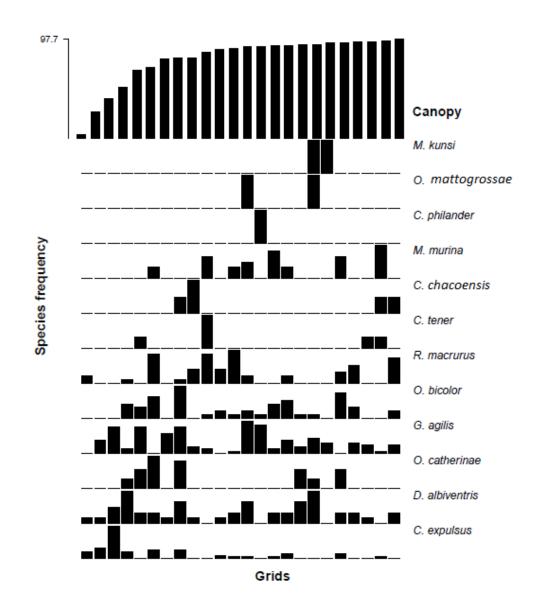


694



696 of fallen logs in fragmented landscape studied.

697

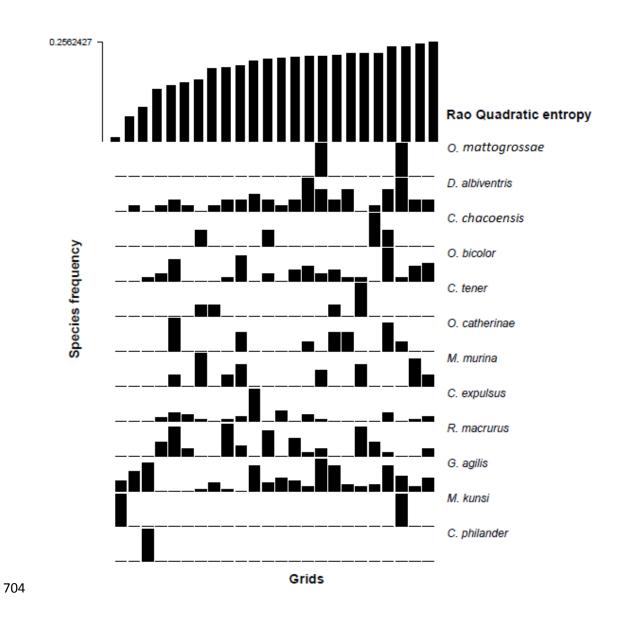


699

700 S2 Fig. Direct ordination of small mammal species' frequency associated with

701 canopy cover in fragmented landscape studied.

702



705 S3 Fig. Direct ordination of small mammal species' frequency associated with Rao

