

# Biogeographical distributions of trickster animals

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## Abstract

Human language encompasses almost endless potential for meaning and folklore can theoretically incorporate themes beyond time and space. However, actual distributions of the themes are not always universal and their constraints remain unclear. Here, we specifically focused on zoological folklore and aimed to reveal what restricts the distribution of trickster animals in folklore. We applied the biogeographical methodology to 16 taxonomic categories of trickster (455 data) and real (93'090'848 data) animals obtained from large databases. Our analysis revealed that the distribution of trickster animals was restricted by their presence in the vicinity and, more importantly, the presence of their corresponding real animals. Given that the distributions of real animals are restricted by the annual mean temperature and annual precipitation, these climatic conditions indirectly affected the distribution of trickster animals. Our study, applying biogeographical methods to culture, paves the way to a deeper understanding of the interactions between ecology and culture.

# 1 Introduction

The hallmark of human language communication is its expressibility. It can enable us to communicate topics remote in time and space (i.e., displacement, [Hockett and Hockett \(1960\)](#)). Folklore is an aspect of human culture that strongly reflects the expressive characteristics of human language. In theory, folklore can refer to animals unseen by storytellers and even describe imaginary animals that do not exist in the real world ([Blust, 2000](#)). Such fictional features can stimulate our curiosity and explorative tendencies ([Dubourg and Baumard, 2022](#)). However, worlds invented for fiction are not free from cognitive constraints. For example, the cost of a magical spell that violates physical laws is not randomly decided; rather, it is based on actual inferences about the physical world ([McCoy et al., 2019](#)). Similarly, ecological factors can restrict the content of folklore. This study focuses on the ecological factors that restrict the theoretically infinite meaning spaces of folklore.

Researchers have discussed the relationship between cultural and ecological factors for decades. Anthropologists, geographers, and other social science and humanities scholars have argued that natural environments are a major source of cultural diversity ([Collard and Foley, 2002](#); [Orlove, 1980](#)); for example, material cultural artefacts such as hunting tools vary across environments ([Osborn, 1999](#); [Peng and Nobayashi, 2021](#)). In addition, the environment can affect nonmaterial cultures. Recent studies show that climatic and/or ecological factors affect political ideologies ([Conway et al., 2020](#)), individualism and collectivism ([Talhelm et al., 2014](#)), social trust ([Dang and Dang, 2021](#)), belief in moralizing gods ([Snarey, 1996](#); [Botero et al., 2014a](#)), and faith in giant trees ([Nakadai, 2023](#)).

Commonly perceived as a collection of traditional stories that transmit cultural identity among social groups, folklore (detailed definition in [Section 2.1](#)) is an example of nonmaterial cultures affected by the environment. Folklore is also vital in acquiring ecological knowledge of the local environment ([Scalise Sugiyama, 2001](#); [Ceriaco et al., 2011](#); [Smith et al., 2017](#)); for instance, the folk-biological knowledge or locals' understanding of harmful animals ([Scalise Sugiyama, 2006](#)), and the pairing of wild and domestic animals in antagonistic interactions ([Nakawake and Sato, 2019](#)).

Biogeography has, for decades, delved into the determinants of species distribution in nature ([Lomolino et al., 2010](#)). Climate conditions are predominant among the numerous biotic and abiotic factors affecting species distributions. For example, many studies have reported shifts in animal and plant distributions due to climate change ([Feehan et al., 2009](#); [Dyderski et al., 2018](#); [Pacifici et al., 2015](#); [Antão et al., 2022](#)). The concept of biomes, or units of plant assemblages and associated animal species, highlights the importance of climate conditions on species distributions ([Smith and Smith, 2012](#); [Gramond, 2021](#); [Hunter et al., 2021](#)); thus, biomes worldwide are classified based on climate conditions ([Moncrieff et al., 2016](#); [Mucina, 2019](#)). One of the most famous biome concepts was proposed by [Whittaker \(1970\)](#). Whittaker's biome classifies the environment into nine (plus one as an outlier) biomes based on annual mean temperature and annual precipitation.

Do animal distributions in folklore reflect the climatic conditions and distributions of real animals? This non-trivial question remains under-researched. Theoretically, folklore can contain any creatures regardless of the local environment due to the expressibility of human language. For example, folklore concerning imaginary

37 animals such as dragons exists worldwide (Blust, 2000; d’Huy, 2013; Jones, 2016), even though such creatures  
 38 do not exist in the real world. Folklore of some real carnivores remains in regions where these animals have  
 39 gone extinct: e.g., bears in Britain (Elms, 1977; O’Regan, 2018) and wolves in Japan (Knight, 1997). The  
 40 distribution of real and fictional animals should be mismatched if motifs of fictional animals are transmitted  
 41 freely across ecological conditions. However, ecological conditions are likely to restrict the animal distribution  
 42 in folklore because folklore contains the ecological knowledge of local environments (Scalise Sugiyama, 2001;  
 43 Ceríaco et al., 2011; Smith et al., 2017).

44 Here, we statistically analyzed databases on folklore, real animals, and climate conditions to find the de-  
 45 terminants of animal distribution in folklore (Fig. 1). We used motifs of animal or zoomorphic tricksters,  
 46 characters performing tricks, because they are stable and worldwide motifs (Berezkin, 2014; Leeming, 2014,  
 47 2022; Pache, 2012); see Section 2.1. for more detailed definition. Berezkin’s collection has accumulated various  
 48 types of folkloristic motifs worldwide, including trickster animals (Berezkin, 2014), with geographic coordinate  
 49 data where folklore was recorded. This provides an ideal opportunity to quantitatively analyze the distribution  
 50 of trickster animals. We hypothesized that (i) climate conditions regulate animal distribution in folklore as in  
 51 nature, and (ii) there is an overlap in the distributions of real and trickster animals in folklore. To test these  
 52 hypotheses, we classified the climate conditions where trickster and/or real animals were sampled into Whit-  
 53 taker’s biomes (Whittaker, 1970). We compared the fractions of the biomes in real and trickster animals and  
 54 found that the distributions of real animals were restricted by climate conditions and that the presence of real  
 55 animals restricted the distributions of trickster animals. In other words, climate conditions indirectly restrict  
 56 the distribution of trickster animals in folklore. These results suggest that ecological factors could restrict the  
 57 contents of folklore or, more broadly, human culture due to human cognitive biases.

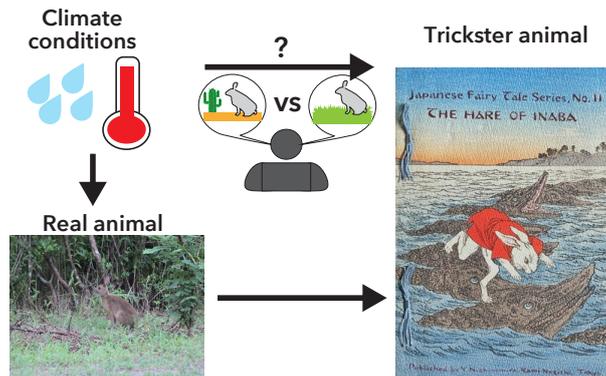


Figure 1: Constraints on the trickster animal distributions

Schematic representations of the manuscript show two environmental conditions: annual mean temperature and annual precipitation. These attributes affect the distribution of real animals that would potentially be represented as tricksters. The distribution of real animals denotes a necessary condition for the presence of corresponding trickster animals. This 1 presents the Japanese hare, *Lepus brachyurus* (Photo by Dr. Abby Darrah <https://www.inaturalist.org/observations/105058298>, CC-BY), and “The Hare of Inaba” (Illustration by Eitaku Kobayashi) as examples of a real and trickster hare, respectively. The image of “The Hare of Inaba” was obtained from the library of the Open University of Japan.

## 58 2 Methods

### 59 2.1 Definitions of folklore, motif, and trickster

60 This subsection describes folklore and details the motif of tricksters in folklore. The term “folklore” can include  
61 material cultures (Brown, 1998) but commonly refers to oral traditions. Bascom (1965) defined folklore as prose  
62 narratives including three categories: folktales, legends, and myths. We use an operational definition of folklore  
63 in this study as any records incorporated in the lifelong work of Dr. Yuri Berezkin, The Thematic Classification  
64 and Areal Distribution of Folklore-Mythological Catalogue (Berezkin, 2015, 2022).

65 The catalog includes more than 3,000 motif indexes developed by Berezkin, who defined motifs as “any  
66 episodes or images retold or described in narratives that are registered in at least in two (although normally in  
67 many more) different traditions” (Berezkin, 2015, p. 37). Berezkin classified motifs into 13 major categories,  
68 labeling them with letters from A to N; among such motifs, themes incorporating tricksters are classified as “M:  
69 ПРИКЛЮЧЕНИЯ III: ПРОДЕЛКИ И ЭПИЗОДЫ (M. Adventures III: Mischief and Episodes; translated by  
70 authors; see <https://www.ruthenia.ru/folklore/berezkin/>.” We used this catalog for two reasons. First,  
71 Berezkin’s catalog includes worldwide folklore (d’Huy et al., 2017), enabling us to compare distributions of  
72 real and trickster animals globally. Second, this catalog provides geographic coordinate data of folklore, which  
73 enables us to compare the distribution of trickster and real animals. This is unique to Berezkin’s catalog because  
74 the Aarne Thompson Uther catalog, which is used in previous studies (Bortolini et al., 2017; Nakawake and  
75 Sato, 2019), does not provide such geographic data. However, there are some drawbacks of Berezkin’s catalog.  
76 Sources of the database are mainly based on literature written in English, Russian, Spanish, German, and  
77 French (Michalopoulos and Xue, 2021). In addition, Berezkin’s catalog does not contain a motif for broader  
78 animal-related tales such as “animal tales,” which the Aarne Thompson Uther catalog contains.

79 Instead of analyzing broad animal folklore, we analyzed the motifs of tricksters because animal or zoomorphic  
80 tricksters are found worldwide and have stable characteristics (Berezkin, 2014). Tricksters are a type of fictional  
81 character that performs tricks and deceptions or exhibits mischievous behaviors (e.g., stealing, cheating). The  
82 trickster’s role is often metaphorically understood: for instance, as “a boundary-crosser” who travels between  
83 or connects two different worlds (Hyde, 2008). Berezkin (2010) defined the trickster as “any personage who  
84 deceives others, acts in a strange way or gets into comical situations but as one who combines two pairs of  
85 opposite characteristics which in the norm are related to different types of actions.”

### 86 2.2 Data collection

87 We compiled data on the distributions of trickster animals from Dr. Berezkin’s world myth database (Berezkin,  
88 2015, 2022), real animals from the Global Biodiversity Information Facility (GBIF) (GBIF.org, 2020), and  
89 climate conditions from WorldClim 2.1 (Fick and Hijmans, 2017). We obtained folklore data via personal  
90 communication with Dr. Yuri Berezkin, downloading it from his database in July 2022. We used the motifs  
91 “Trickster–X” [m29a – m29i] and “Trickster is a(n) X” [m29l –m29y]. The items encased in square brackets

92 show Berezkin’s motif index and  $X$  represents common animal names. We analyzed motifs that satisfied the  
93 following three criteria to proceed with the further analysis:

- 94 1. The scientific names can be estimated from the focal animals’ common names because scientific names  
95 were needed to obtain the animal distribution from GBIF,
- 96 2. Grouping multiple animal names together was allowed when it was taxonomically reasonable, and
- 97 3. Assembling multiple animals was avoided when their distribution was known to be geographically distinct;  
98 otherwise the distribution of trickster and real animals would be biased to overlap more.

99 In this study, the following 16 categories satisfied the above criteria and were analyzed: anteater [m29qq], badger  
100 [m29x1], hawk [m29i], mink [m29d], mouse [m29n], opossum [m29l], owl [m29h], porcupine [m29r], rabbit/hare  
101 [m29g], raccoon [m29q], rat [m29m], raven/crow [m29a], skunk [m29c], spider [m29p], and wren [m29y]. We  
102 removed six motifs from the analysis because they did not satisfy either of the three criteria: (i) monkeys [m29o],  
103 (ii) water birds [m29j], (iii) foxes, coyotes, or jackals [m29b], (iv) felines (jaguars, ocelots, or pumas) [m29w],  
104 (v) small ungulates [m29v], and (vi) turtles, toads, or frogs [m29k].

105 For example, the types of animals to be included in water birds [m29j] and small ungulates [m29v] were  
106 unclear (not satisfying criterion 1), and we could not specify the scientific names of species corresponding to  
107 these animals. Similarly, we could not proceed with the analysis of monkeys [m29o] because what “monkey”  
108 includes changes over time, and this category can be vaguely used (e.g., whether monkeys include apes or  
109 not); see Oxford English Dictionary for details: [https://www.oed.com/dictionary/monkey\\_n?tab=meaning\\_](https://www.oed.com/dictionary/monkey_n?tab=meaning_and_use&tl=true#36269827)  
110 [and\\_use&tl=true#36269827](https://www.oed.com/dictionary/monkey_n?tab=meaning_and_use&tl=true#36269827). Grouping turtles, toads, and frogs together [m29k] is biologically unreasonable  
111 (not satisfying criterion 2) as turtles are reptiles while toads and frogs are amphibians. Foxes, coyotes, or  
112 jackals [m29b] should be subdivided because the previous study shows that the geographic distribution of their  
113 corresponding trickster animals does not overlap (not satisfying criterion 3) (Berezkin, 2014). Felines [m29w]  
114 include many species whose geographic distributions are distinct (not satisfying criterion 3) (O’Brien et al.,  
115 2008). Because the details of these folklores were unavailable, we could not subdivide these data and removed  
116 them from further analyses. The amount of data remaining for each trickster animal ranged from 6 to 190 (a  
117 total of 455 pieces of data) depending on the category.

118 We used Wikipedia to assign the scientific names of the corresponding real animals for each trickster animal.  
119 We confirmed whether these suggested scientific names matched the common names of the animals by accessing  
120 the National Center for Biotechnology Information and the Encyclopedia of Life using the `sci2comm()` function  
121 in the `taxize` library (Chamberlain et al., 2013) version 0.9.98 in R (version 4.2.1). Four scientific names (two  
122 ground squirrels: *Geosciurus* and *Euxerus*, and two badgers: *Arctonyx hoevenii* and *Melogale subaurantiaca*)  
123 did not appear on either database, and we removed these species from further analysis (see also supplementary  
124 data). The distributions of the real animals were collected from GBIF using the `occ_download` function in  
125 the `rgbif` library version 3.7.3 (Chamberlain et al., 2022) in R. The coordinate data were cleaned using the  
126 `clean_coordinates` function of the `CoordinateCleaner` library (Zizka et al., 2019) with tests of capitals, centroids,

127 gbif, institutions, and zeros. After data cleaning, the data of real animals varied from 5'400 to 50'000'000 (a  
128 total of 93'090'848 pieces of data) depending on the animal category.

129 The intensity of data collection relating to tricksters and real animals would probably differ across species  
130 and locations. Therefore, we converted the coordinate data into hex grid indices using the `geo_to_h3` function  
131 in `h3` package version 3.7.4 (Uber Technologies Inc., 2018) of Python 3 (version 3.8.13). The resolution of the  
132 hex grid is crucial in our analysis. This parameter determines the number of grids where the tricksters and/or  
133 real animals exist. Because the number of trickster data pieces is small, enhancing the resolution parameter  
134 would increase the statistical power. Meanwhile, the climate conditions may be unavailable with the higher  
135 resolution, and the computational costs of the analyses increase over the resolution. We set the resolution of  
136 the hex grids = 1, generating 842 grids across the world map, because the number of grids is larger than the  
137 number of trickster data pieces and because the climate data (see below) are assigned to almost all grids. Table  
138 S1 shows that the number of grids that the presence of the tricksters were reported did little change when the  
139 resolution parameter is two or higher. In the supplementary data, we show the results with the resolution of  
140 the hex grid = 2 (5882 grids across the world), but these analyses show qualitatively similar results with the  
141 main text (Tables S2, S3, and S4). We did not consider the number of reports per grid in this manuscript; we  
142 used only the presence data of the tricksters and real animals in each grid to minimize the effect of sampling  
143 biases across species and space. After the data conversion, we obtained 257 pieces of presence data of trickster  
144 animals on the hex grids and 3'413 pieces of presence data of real animals on the hex grids.

145 The climate data were assigned to each hex grid after the coordinates of tricksters and real animals had been  
146 converted. We retrieved the annual mean temperature and annual precipitation of the center point of each grid  
147 from WorldClim 2.1 (Fick and Hijmans, 2017) using the `latlon-utils` package version 0.07 (Sommer, 2022) in  
148 Python 3. We selected data on these two climate conditions because they classify environment into Whittaker's  
149 biomes (Whittaker, 1970). If the annual mean temperature and/or annual precipitation were unavailable (for  
150 example, when a center point of a grid existed on an ocean), we estimated the two groups of environmental  
151 data from the means at the coordinates inside the grid at which real animals were reported. We grouped the  
152 data into biome classes using the `plotbiomes` library (Stefan and Levin, 2022) in R.

## 153 2.3 Statistical analyses

154 We first investigated the fractions of Whittaker's biome classes. For each animal category, we compared the  
155 fractions of the biome classes between the tricksters and corresponding real animals. Furthermore, we compared  
156 the fractions of the biome classes with a null model generated by the hex grids and corresponding environmental  
157 conditions where at least one of the real animals in our analysis was reported. This null model represents the  
158 fractions of the biome classes in terrestrial areas. We used the chi-squared test in R to compare the fractions  
159 of the biome classes. We corrected the obtained p-values using the false discovery rate (FDR) method with the  
160 `p.adjust` function.

161 We then investigated whether the presence of tricksters in each grid was limited by the presence of the

162 corresponding real animals. We calculated the conditional probabilities that the corresponding real animals  
163 were reported in a grid within which the focal animals appeared as tricksters in folklore. This conditional  
164 probability represents whether the corresponding real animals regulate the presence of trickster animals. A  
165 very low conditional probability would imply that trickster folklore could be transmitted to areas in which the  
166 locals were unfamiliar with the focal animals. Conversely, a high conditional probability would suggest that the  
167 presence of real animals was a necessary condition for the presence of trickster animals in the folklore. Notably,  
168 this conditional probability did not intend to show the predictability of the presence of trickster animals, which  
169 is beyond the scope of this study.

170 Next, we performed a permutation test to determine whether the distribution of each trickster animal was  
171 clogged. The above analysis indicated that the presence of the corresponding real animals was necessary for the  
172 presence of a trickster in the folklore (Fig. 3). Therefore, the null hypothesis was postulated—a focal animal  
173 appears as a trickster where the corresponding real animals are observed. We compared the median distance  
174 between the hex grids where the focal animals were reported as tricksters and the median of the simulated  
175 distances under the null hypothesis. The distributions of trickster animals under the null hypothesis were  
176 generated by randomly selecting the hex grids within which the corresponding real animals existed, and the  
177 number of selected grids was identical to the number of grids that the focal trickster animals were reported. We  
178 generated 10'000 such distributions for each animal and obtained the probability distributions of the median  
179 distances according to the null hypothesis, which enabled us to calculate p-values. The attained p-values were  
180 corrected by the FDR method using the `multitest.fdr` function in the `statsmodels` library (Seabold  
181 and Perktold, 2010) in Python 3.

## 182 3 Results

### 183 3.1 Environmental constraints on animal distributions

184 We investigated the effects exerted by climate conditions on the distributions of real and trickster animals  
185 (Fig. 2). We classified climate conditions into nine groups (and one as an outlier) as per Whittaker's biome  
186 classes (Whittaker, 1970) and compared the fractions of the biome classes between each category of animal and  
187 terrestrial areas (i.e., the null model). The left column of Table 1 shows that the distributions of 12 of the  
188 16 real animals differ from the null model, suggesting that annual mean temperature and annual precipitation  
189 restrict the distribution of many animals. The exceptional animals (i.e., hawk, owl, rabbit or hare, and spider)  
190 were found on all continents except Antarctica. In contrast, only four animals (mink, opossum, raven or crow,  
191 and skunk) differed in the fractions of biome classes between the tricksters and the null model (the middle  
192 column of Table 1). Trickster minks were found in temperate seasonal forests, opossums were noted in tropical  
193 seasonal forests/savannas, ravens or crows were observed in the tundra, boreal forests, temperate seasonal forests,  
194 or tropical seasonal forests/savannas, and skunks were seen in boreal forests or temperate seasonal forests.  
195 These analyses provide evidence that annual mean temperature and annual precipitation restrict real animal

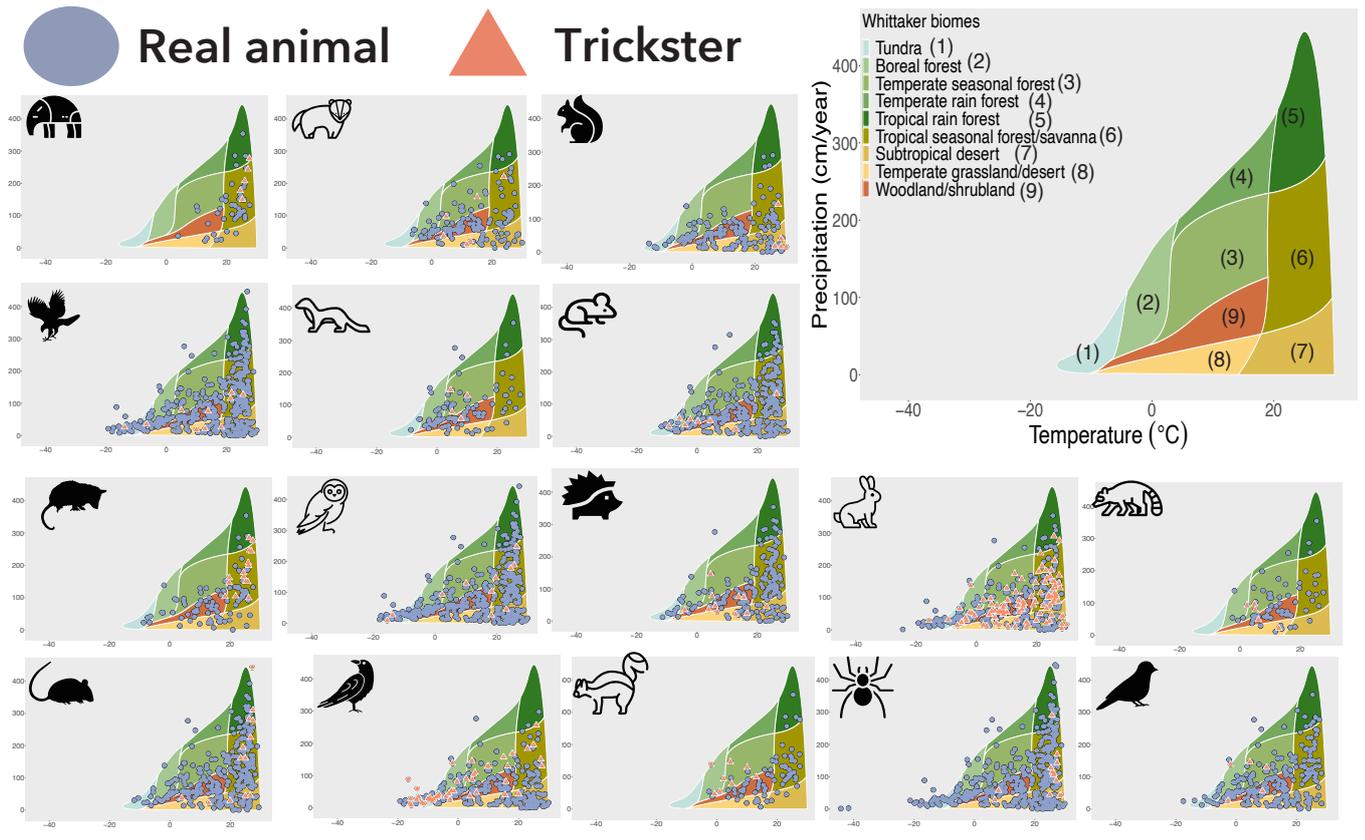


Figure 2: The distribution of real and trickster animals in Whittaker's biome

The distributions of 16 real and trickster animals (shown by icons) are shown on Whittaker's biome (top right). The blue circles and the orange triangles in Whittaker's biome depict the climate conditions of the regions where the real animals and tricksters were reported, respectively. The background colours and the numbers in parentheses represent the biome classes (see the bottom right panel). The enlarged figures are available in Figs. S1-S16.

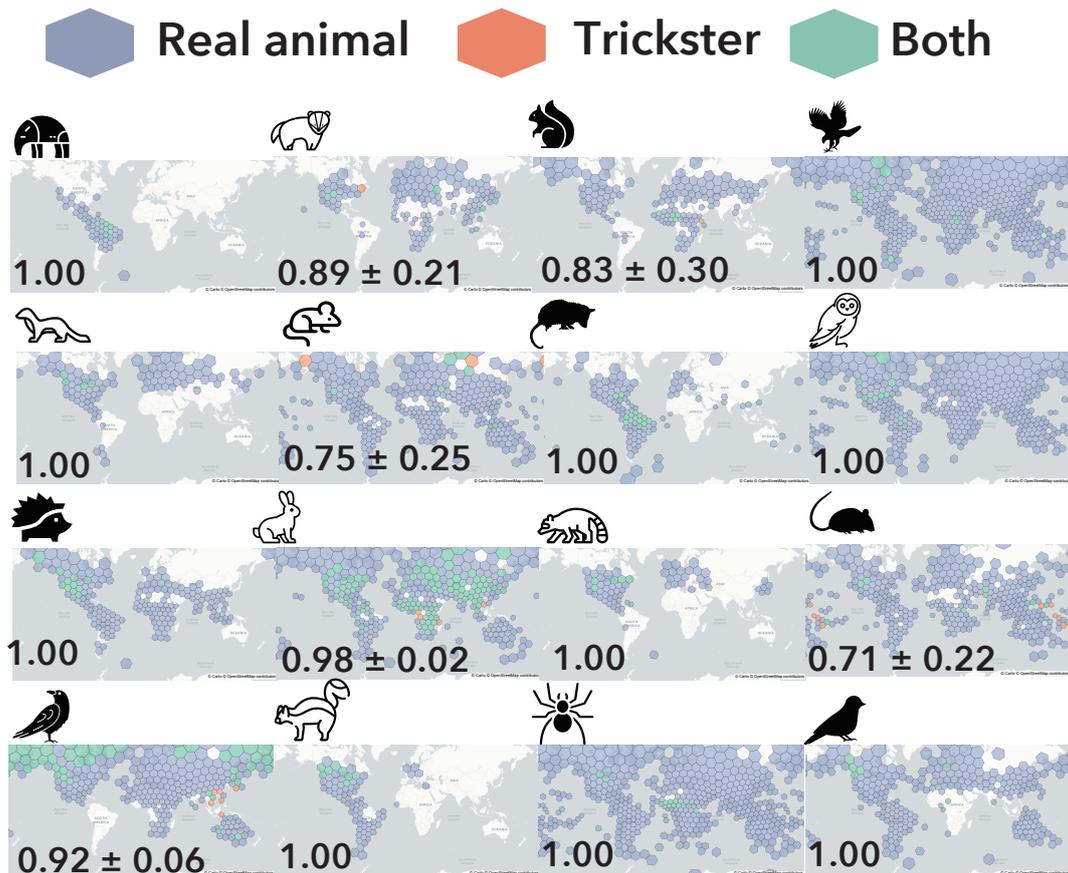


Figure 3: The distribution of real and trickster animals on the world map

The distributions of 16 real and trickster animals (shown by icons) are shown on the world map (top). The blue, orange, and green hex grids respectively represent where only the real animals, only the trickster animals, or both versions were reported, respectively. The numbers at the bottom left indicate the conditional probabilities that the corresponding real animals existed in the grid where the trickster animals were reported, and their 95% confidence intervals. The enlarged figures are available in Figs. S1–S16.

196 distributions; however, such environmental constraints are less evident on trickster animal distributions. This  
 197 may, however, be due to differences in the amounts of data (see Section 2.2). The quantity of trickster-related  
 198 data with the grid resolution parameter = 1 may be too small (between 4 to 99 pieces of data depending on the  
 199 animal category, see Table S1) in comparison to the number of biome classes (totaling 10); thus, the statistical  
 200 power may not be large enough. Indeed, this result was sensitive to the resolution parameter; increasing the  
 201 resolution of the grids shows that fractions of tricksters’ biomes are different from the null model in 12 animal  
 202 categories (the middle column of Table S2) because of the increase in the amount of trickster’s data.

Table 1: P-values in chi-squared test to compare the frequencies of the biome classes

Category	Real vs Null	Trickster vs Null	Real vs Trickster
Anteater	$1.46 \times 10^{-4}$ ✓	$5.00 \times 10^{-1}$	$8.26 \times 10^{-1}$
Badger	$2.13 \times 10^{-5}$ ✓	$1.02 \times 10^{-1}$	$5.47 \times 10^{-1}$
Ground squirrel	$2.09 \times 10^{-7}$ ✓	$5.00 \times 10^{-1}$	$2.13 \times 10^{-1}$
Hawk	$9.96 \times 10^{-1}$	$6.29 \times 10^{-1}$	$7.55 \times 10^{-1}$
Mink	$2.59 \times 10^{-9}$ ✓	$4.08 \times 10^{-2}$ ✓	$5.72 \times 10^{-1}$
Mouse	$1.77 \times 10^{-2}$ ✓	$7.11 \times 10^{-2}$	$9.78 \times 10^{-4}$ ✓
Opossum	$1.07 \times 10^{-2}$ ✓	$4.08 \times 10^{-2}$ ✓	$1.80 \times 10^{-1}$
Owl	$9.96 \times 10^{-1}$	$8.47 \times 10^{-1}$	$7.55 \times 10^{-1}$
Porcupine	$3.38 \times 10^{-2}$ ✓	$2.45 \times 10^{-1}$	$2.18 \times 10^{-1}$
Rabbit/Hare	$8.00 \times 10^{-2}$	$7.99 \times 10^{-2}$	$2.92 \times 10^{-1}$
Raccoon	$3.56 \times 10^{-7}$ ✓	$3.45 \times 10^{-1}$	$7.69 \times 10^{-1}$
Rat	$2.99 \times 10^{-4}$ ✓	$5.00 \times 10^{-1}$	$5.47 \times 10^{-1}$
Raven/Crow	$2.49 \times 10^{-8}$ ✓	$1.55 \times 10^{-7}$ ✓	$1.81 \times 10^{-5}$ ✓
Skunk	$6.45 \times 10^{-3}$ ✓	$4.08 \times 10^{-2}$ ✓	$4.51 \times 10^{-4}$ ✓
Spider	$9.96 \times 10^{-1}$	$6.29 \times 10^{-1}$	$7.55 \times 10^{-1}$
Wren	$8.84 \times 10^{-7}$ ✓	$3.40 \times 10^{-1}$	$5.44 \times 10^{-1}$

✓ represents p-value after FDR correction < 0.05.

### 203 3.2 Ecological constraints on animal tricksters

204 Next, we determined whether the trickster animals were freely distributed across the world or whether their  
 205 presence was restricted by the presence of their corresponding real animals. For this purpose, we calculated  
 206 the conditional probability that a corresponding real animal existed in the region where the trickster animal  
 207 appeared in local folklore. The values in Fig. 3 show that the conditional probabilities of 14 animals were  
 208 greater than 80%, suggesting that the presence of real animals is an almost necessary condition for the presence  
 209 of trickster animals. Qualitatively similar results were obtained when we increased the resolution of the hex  
 210 grids (Table S3). As the real animal distributions were restricted by the two climate conditions, we concluded  
 211 that these conditions indirectly restricted the distribution of the trickster animals. Further constraints were  
 212 unclear because only three trickster animals (i.e., mouse, raven or crow, and skunk) differed in the fractions of  
 213 the biome classes from their corresponding real animals (the right column of Table 1). This may again reflect  
 214 a small statistical power due to the small pieces of the tricksters’ data; increasing their data via enhancing the  
 215 grids’ resolution revealed that the biome fractions between real and trickster animals significantly differ in ten  
 216 animals (the right column of Table S2).

217 Mice and rats showed exceptionally lower conditional probabilities than the other animals. Although these  
 218 species appeared in certain regions where only tricksters were observed, such areas were surrounded by the

219 regions in which real mice and rats were seen (i.e., the orange areas surrounded by blue or green areas on the  
220 world maps in Fig. 3).

### 221 3.3 Constraints by neighbour tricksters

222 We also investigated whether the presence of trickster animals was affected by other tricksters in the neigh-  
223 bourhoods (i.e., surrounding grids). The distance between societies with identical trickster animals would be  
224 shorter if these folklores were culturally transmitted from one to another than if these trickster animals were  
225 independently created in each society with a certain probability. Clusters of trickster animals are displayed on  
226 the world maps Fig.3. Potential restriction of trickster distribution within a part of biomes (the right column  
227 of Table S2) may reflect the fact that closer areas have similar climate conditions. The permutation test also  
228 revealed that the distance between the grids where trickster animals existed was shorter for 13 animals than  
229 the distance between randomly chosen grids in which the corresponding real animals existed (Fig. 4). These  
230 animals and the p-value calculated after FDR correction are noted on Table 2. Increasing the resolution of the  
231 grid did not change the results of the permutation tests (Table S4) Therefore, the tricksters of a focal animal  
232 were positively affected by the presence of other tricksters in the vicinity.

Table 2: P-values in the permutation test

Anteater	$9.58 \times 10^{-3}$	✓
Badger	$7.74 \times 10^{-1}$	
Hawk	$6.12 \times 10^{-1}$	
Mink	$1.45 \times 10^{-2}$	✓
Mouse	$7.06 \times 10^{-3}$	✓
Opossum	$9.85 \times 10^{-11}$	✓
Owl	$1.23 \times 10^{-4}$	✓
Porcupine	$1.72 \times 10^{-21}$	✓
Rabbit/Hare	$1.03 \times 10^{-6}$	✓
Raccoon	$3.69 \times 10^{-2}$	✓
Rat	$4.99 \times 10^{-1}$	
Raven/Crow	$4.42 \times 10^{-10}$	✓
Skunk	$1.28 \times 10^{-4}$	✓
Spider	$6.50 \times 10^{-59}$	✓
Wren	$1.23 \times 10^{-4}$	✓.

✓represents p-value < 0.05 after FDR correction.

## 233 4 Discussion

234 Folklore is one of the human cultures that have the most enriched records, and the diffusion of folklore has  
235 been investigated as an example of human cultural evolution (Graça da Silva and Tehrani, 2016; Bortolini  
236 et al., 2017). Because human imagination is boundless and human languages are almost unlimited in terms of  
237 expression (Hockett and Hockett, 1960), stories can contain creatures never witnessed by their tellers. Hence,  
238 fictional creatures in folklore could theoretically be shared worldwide via cultural transmission. This study,  
239 however, demonstrates that the presence of real animals is almost a prerequisite for trickster animals to appear.  
240 In other words, ecological and climatic conditions have dominant effects on contents in folklore (Scalise Sugiyama,

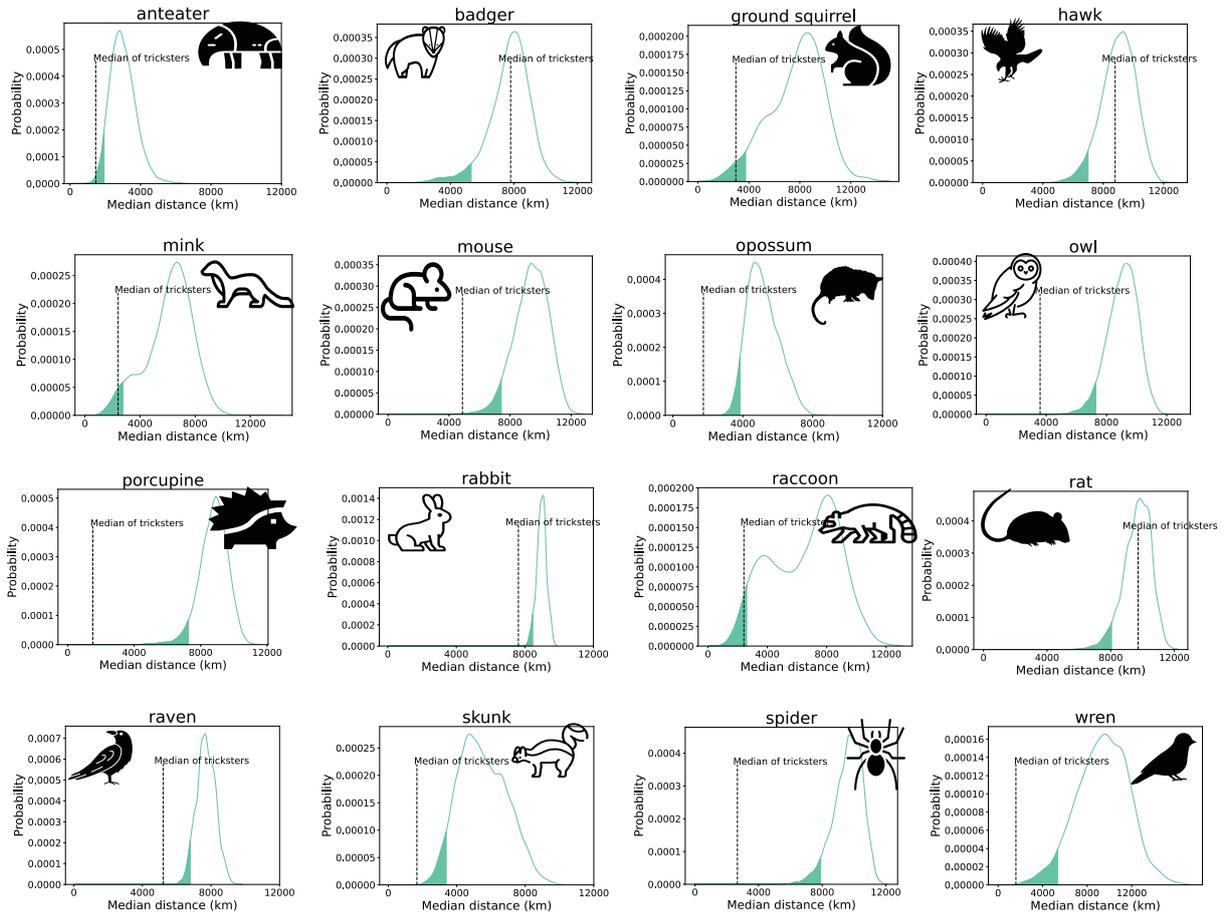


Figure 4: Permutation test of the distances on the world map

In the null model, the trickster animals were positioned randomly on the grid in places where the corresponding real animals were reported. For each animal, we determined the locations in which the corresponding tricksters were more densely distributed. The dashed line in every panel represents the median distance between the tricksters in the data; the curve represents the probability distribution of the median distance per the null model; and the shaded areas indicate the lowest five percent values of the distribution. The p-values after FDR correction have been noted Table 2.

241 2001; Ceriaco et al., 2011; Smith et al., 2017), as on other human culture (Collard and Foley, 2002; Orlove,  
242 1980; Osborn, 1999; Peng and Nobayashi, 2021; Conway et al., 2020; Talhelm et al., 2014; Dang and Dang,  
243 2021; Snarey, 1996; Botero et al., 2014b; Nakadai, 2023).

244 This study applied a biogeographical methodology to demonstrate how certain cultural notions (in this  
245 instance, folk motifs) are limited by local ecological factors. The folklore of societies is unlikely to include  
246 focal trickster animals if the corresponding real animals were not reported there. Trickster mice and rats  
247 were exceptions; we could not, however, conclude whether the real mice and rats were missing because our data  
248 indicate only the presence, but not the absence, of the animals. For the rest animals, the distributions of trickster  
249 and real animals overlapped. The annual mean temperature and annual precipitation affect the distribution of  
250 many real animals. Hence, these climate conditions indirectly restrict the distributions of trickster animals in  
251 folklore (Fig. 1).

252 Fig. 4 shows that the distance between reported trickster animals was closer than that when trickster  
253 animals were randomly distributed to where the corresponding real animals existed. Although such patterns  
254 would occur if the trickster folklore was culturally transmitted from the neighborhood, other mechanisms can  
255 also produce patterns. For example, the geographically biased sampling of folklore can generate similar patterns.  
256 Alternatively, environmental conditions that Whittaker's biome does not include may affect the distribution of  
257 tricksters. In this case, closer areas may have more similar environmental conditions. To analyze whether closer  
258 trickster folklore was culturally transmitted or not, one potential future research direction is to reconstruct the  
259 dynamics of folklore diffusion by, for example, cultural phylogenetics (Tehrani, 2013; Martini, 2020).

260 Once the diffusion of folklore is reconstructed, this would pave the way to investigate the mechanisms to  
261 generate the patterns (Figs. 2 and 3) observed in this study. Humans tend to focus on familiar informational  
262 content and reproduce stories as per content or schematic frameworks (i.e., schema) that they already know  
263 (Lyons and Kashima, 2006; Hunzaker, 2016). Previous experiments have shown that cognitive biases shape  
264 folklore in certain directions (Lyons and Kashima, 2006; Hunzaker, 2016; Stubbersfield, 2022). Such cognitive  
265 or behavioural processes may shape folklore incorporating trickster animals whose corresponding real animals  
266 were familiar to locals. If this is the case, we can hypothesize that the presence of real animals enhances the  
267 creation, adaptation, or maintenance of corresponding trickster animals. The extinction rate of the tricksters,  
268 on the other hand, might be independent of the presence/absence of real animals because some carnivores'  
269 tales remain in the area where the corresponding real animals have gone extinct (Elms, 1977; O'Regan, 2018;  
270 Knight, 1997). Although cultural extinction has been analyzed theoretically and empirically (Kobayashi et al.,  
271 2021; Zhang and Mace, 2021), Berezkin's folklore database is not suitable for such analyses because dynamics  
272 of the presence/absence of folklore in each area are not available. Once the time series data of folklore and  
273 real animals are available, one can test whether the presence of real animals affects the creation/acceptance or  
274 extinction rates of trickster animals by comparing the empirical distributions of real and trickster animals with  
275 a null model that does incorporate the presence/absence of real animals. Such a null model can be built based  
276 on the dual inheritance theory that allows mismatches between environments and cultural traits (Richerson and  
277 Boyd, 2006).

278 One limitation of this study is that tricksters are subsets of animal folklore. Broader animal folklores can  
279 be analyzed by the motif of “animal tales” in Aarne Thompson Uther catalog (Nakawake and Sato, 2019),  
280 although it does not provide the geographic coordinate information of folklore. There are overlaps between  
281 trickster animals and animals in the motif of “animal tales,” but some animals that frequently appeared only  
282 in the motif of “animal tales” (see Supplementary Table S1 in Nakawake and Sato (2019)) were not reported as  
283 tricksters. Future studies are needed to investigate whether the natural environments restrict the distribution  
284 of broader animal tales or not. More generally, future research could expand our framework to broader fictional  
285 creatures to investigate whether contents of folklore are, in general, restricted by local environments. For  
286 example, folklore related to dragons, water-related chimeric creatures whose bodies are partially that of snakes,  
287 is described in all continents (Blust, 2000; d’Huy, 2013; Jones, 2016). Blust (2000) argues that dragons were  
288 inspired by the rainbow, a natural phenomenon worldwide. This argument would be supported by investigations  
289 of climate conditions to find correlations between dragon-related folklore and the occurrence of rainbows. One  
290 obstacle of such research would be how to determine the pairs of supernatural creatures with the motifs they  
291 are based on because the ontology of supernatural creatures can vary among literature.

292 The recent increase in quantitative analyses of cultural resources has advanced our understanding of human  
293 cultures by incorporating theories and methodologies employed in evolutionary biology (Tehrani, 2013; Mar-  
294 tini, 2020). Our investigation incorporates biogeographical theories and methods to explore the links between  
295 folkloristic traditions and local ecological conditions. We believe that biogeographical concepts, particularly  
296 Whittaker’s biome scheme, would enrich our understanding of the relationships between human culture and  
297 ecology. Future studies could also apply ecological approaches to move from investigating restrictions to pre-  
298 dicting cultural distribution. Ecologists have developed statistical methods to predict the distribution of species.  
299 However, these methodologies can also apply to fictional creatures (Warren et al., 2021) and institutions (Ai  
300 et al., 2022). Such analyses employ aspects such as climate conditions, the distribution of other species (poten-  
301 tially including cultures and institutions), and their interactions (Pollock et al., 2014). Further, ecologists have  
302 investigated the determiners of biodiversity and temporal stability of systems (May, 1972; Shmida and Wilson,  
303 1985; Landi et al., 2018), which would be applicable to investigate the stability and diversity of human culture.  
304 Collaboration with ecologists and evolutionary biologists would be promising to deepen the understanding of  
305 human culture.

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## Data availability

The original data on folklore is available from Dr. Yuri Berezkin at Department of Anthropology, the European University at Saint Petersburg. The codes and derived data used in this manuscript are available from <https://github.com/ShotaSHIBASAKI/DistributionTrickSter>.

## Author contributions

S.S., R.N., and Y.N. conceived the research, S.S., R.N., and Y.N. compiled data, S.S. performed the statistical analysis, and S.S. wrote the first draft. All authors revised the manuscript and approved the final draft for publication.

## Conflict of interest

The authors declare no conflict of interest.

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Table S1: The number of grids tricksters are reported over the resolution parameter

Category\Resolution	0	1	2	3	4	5
Anteater	4	5	8	9	10	10
Badger	6	6	9	9	9	9
Grand squirrel	3	5	5	6	6	6
Hawk	7	10	12	13	13	13
Mink	3	4	6	8	9	9
Mouse	6	8	12	13	13	13
Opossum	7	15	23	23	25	25
Owl	8	11	14	14	14	14
Porcupine	5	11	15	18	18	18
Rabbit/Hare	38	99	168	187	189	190
Raccon	4	5	8	9	9	9
Rat	10	14	17	17	17	17
Rave/Crow	21	44	74	78	79	79
Skunk	4	8	10	10	12	12
Spider	6	13	20	23	23	23
Wren	3	5	7	8	9	9

Table S2: P-values in chi-squared test to compare the frequencies of the biome classes with higher resolution data

Category	Real vs Null	Trickster vs Null	Real vs Trickster
Anteater	$1.15 \times 10^{-47}$ ✓	$3.26 \times 10^{-3}$ ✓	$9.04 \times 10^{-1}$
Badger	$1.42 \times 10^{-27}$ ✓	$4.48 \times 10^{-1}$	$5.72 \times 10^{-4}$ ✓
Ground squirrel	$5.28 \times 10^{-22}$ ✓	$4.48 \times 10^{-1}$	$2.27 \times 10^{-1}$
Hawk	$2.64 \times 10^{-12}$ ✓	$5.36 \times 10^{-1}$	$8.92 \times 10^{-1}$
Mink	$1.02 \times 10^{-90}$ ✓	$2.09 \times 10^{-3}$ ✓	$3.06 \times 10^{-5}$ ✓
Mouse	$5.22 \times 10^{-9}$ ✓	$1.34 \times 10^{-6}$ ✓	$1.93 \times 10^{-15}$ ✓
Opossum	$6.48 \times 10^{-17}$ ✓	$3.72 \times 10^{-12}$ ✓	$2.86 \times 10^{-8}$ ✓
Owl	$4.84 \times 10^{-7}$ ✓	$2.61 \times 10^{-1}$	$3.40 \times 10^{-1}$
Porcupine	$6.51 \times 10^{-15}$ ✓	$2.64 \times 10^{-3}$ ✓	$2.35 \times 10^{-5}$ ✓
Rabbit/Hare	$5.81 \times 10^{-10}$ ✓	$1.46 \times 10^{-2}$ ✓	$1.92 \times 10^{-8}$ ✓
Raccoon	$3.27 \times 10^{-17}$ ✓	$1.57 \times 10^{-2}$ ✓	$1.10 \times 10^{-1}$
Rat	$4.68 \times 10^{-22}$ ✓	$1.93 \times 10^{-10}$ ✓	$1.82 \times 10^{-16}$ ✓
Raven/Crow	$2.16 \times 10^{-73}$ ✓	$5.07 \times 10^{-50}$ ✓	$3.35 \times 10^{-20}$ ✓
Skunk	$8.63 \times 10^{-9}$ ✓	$6.37 \times 10^{-4}$ ✓	$3.23 \times 10^{-40}$ ✓
Spider	$2.32 \times 10^{-36}$ ✓	$1.95 \times 10^{-2}$ ✓	$8.35 \times 10^{-2}$
Wren	$2.31 \times 10^{-9}$ ✓	$8.79 \times 10^{-5}$ ✓	$7.96 \times 10^{-5}$ ✓

✓ represents p-value after FDR correction  $< 0.05$ .

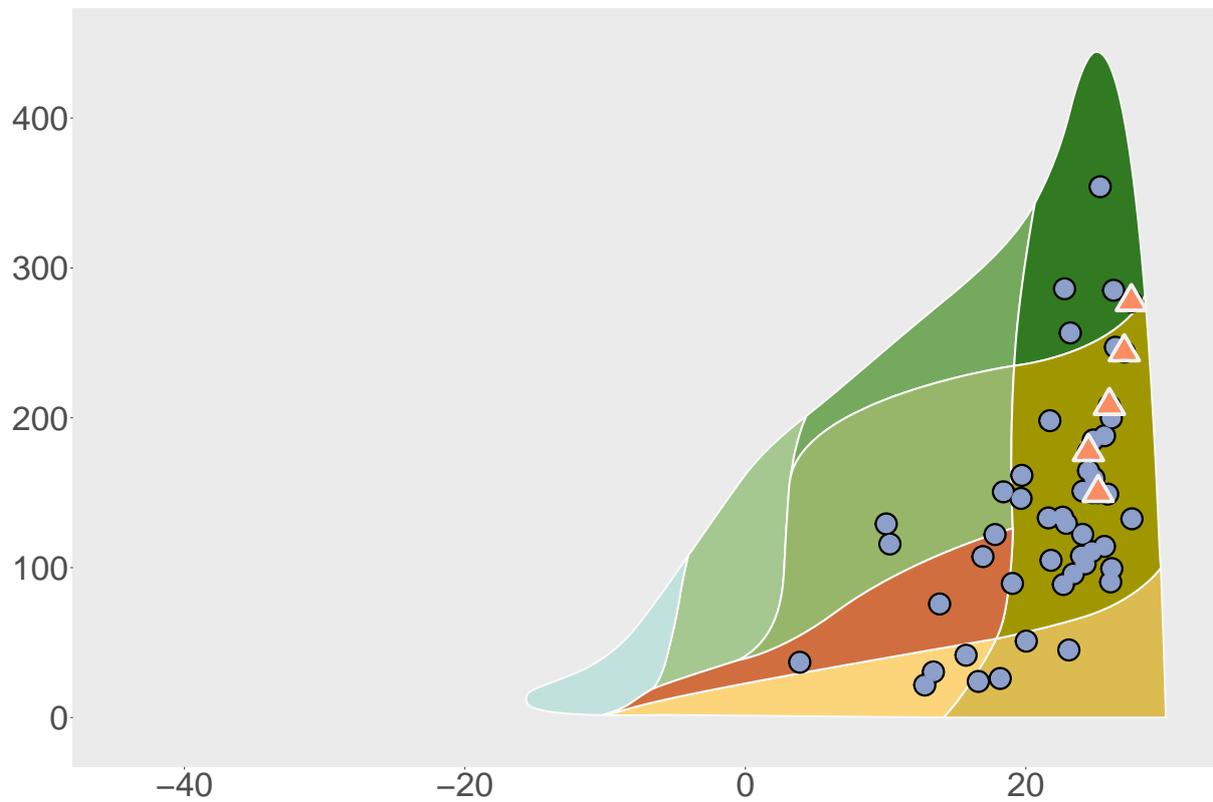
Table S3: The conditional probability that the corresponding animals existed in the grid where the trickster animals were reported, with the resolution parameter = 2

Category	Probability	95%CI
Anteater	0.70	0.28
Badger	0.89	0.21
Grand squirrel	0.83	0.30
Hawk	1.00	0.00
Mink	1.00	0.00
Mouse	0.33	0.27
Opossum	0.92	0.11
Owl	0.93	0.13
Porcupine	1.00	0.00
Rabbit/Hare	0.74	0.06
Raccon	1.00	0.00
Rat	0.59	0.23
Rave/Crow	0.84	0.08
Skunk	0.83	0.21
Spider	0.91	0.12
Wren	1.00	0.00

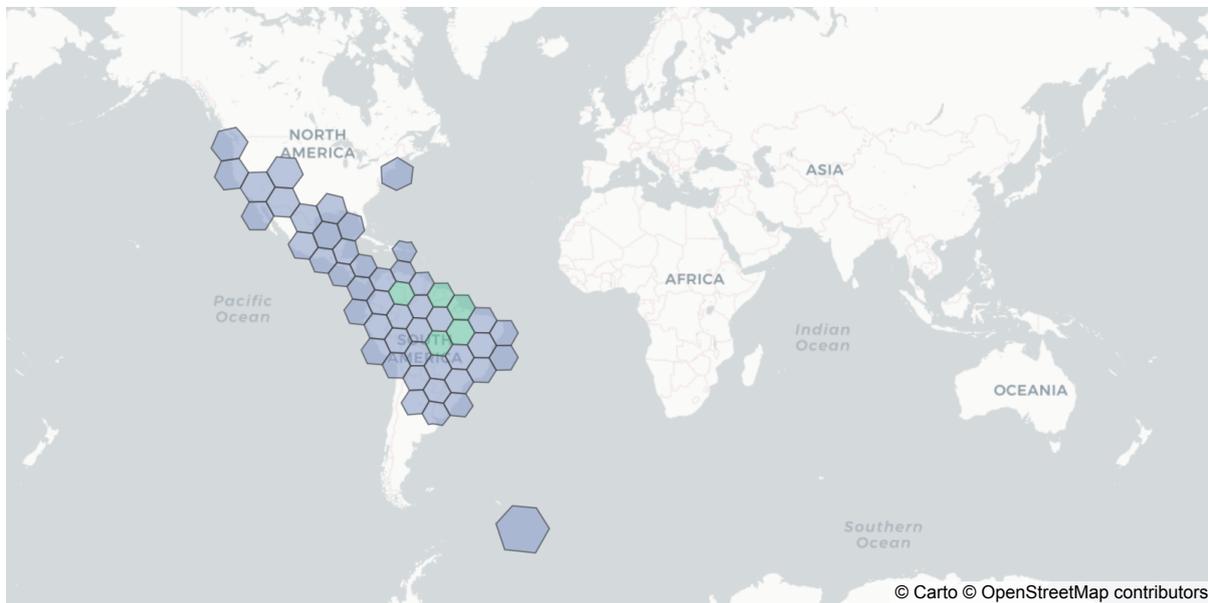
CI represents the conditional interval.

Table S4: Permutation tests at resolution = 2

Category	P-values (after FDR)	Significance
Anteater	$6.48 \times 10^{-3}$	✓
Badger	$4.71 \times 10^{-1}$	
Ground squirrel	$2.65 \times 10^{-2}$	✓
Hawk	$4.05 \times 10^{-1}$	
Mink	$3.01 \times 10^{-2}$	✓
Mouse	$4.08 \times 10^{-4}$	✓
Opossum	$1.01 \times 10^{-12}$	✓
Owl	$8.42 \times 10^{-17}$	✓
Porcupine	$6.57 \times 10^{-40}$	✓
Rabbit	$7.52 \times 10^{-6}$	✓
Racoon	$1.46 \times 10^{-2}$	✓
Rat	$3.67 \times 10^{-1}$	
Raven	$2.86 \times 10^{-112}$	✓
Skunk	$1.72 \times 10^{-3}$	✓
Spider	$3.05 \times 10^{-176}$	✓
Wren	$2.39 \times 10^{-7}$	✓



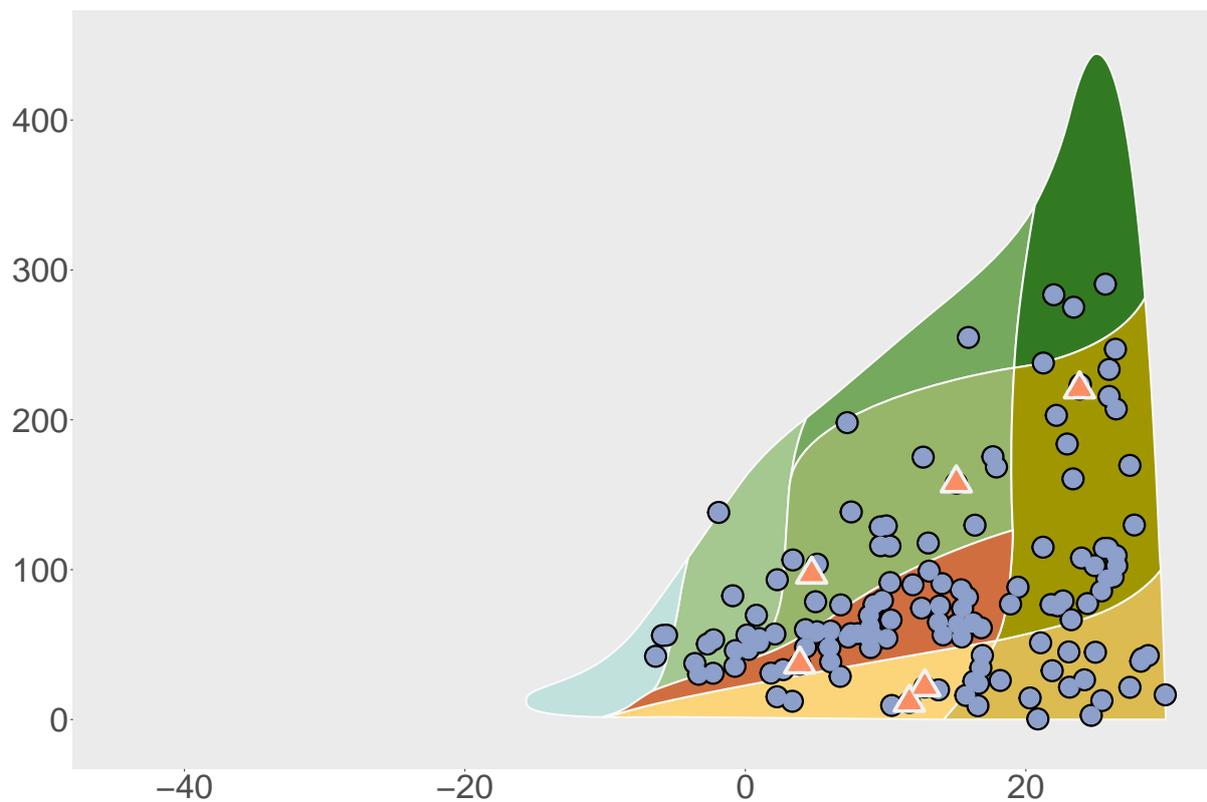
(a) Biome



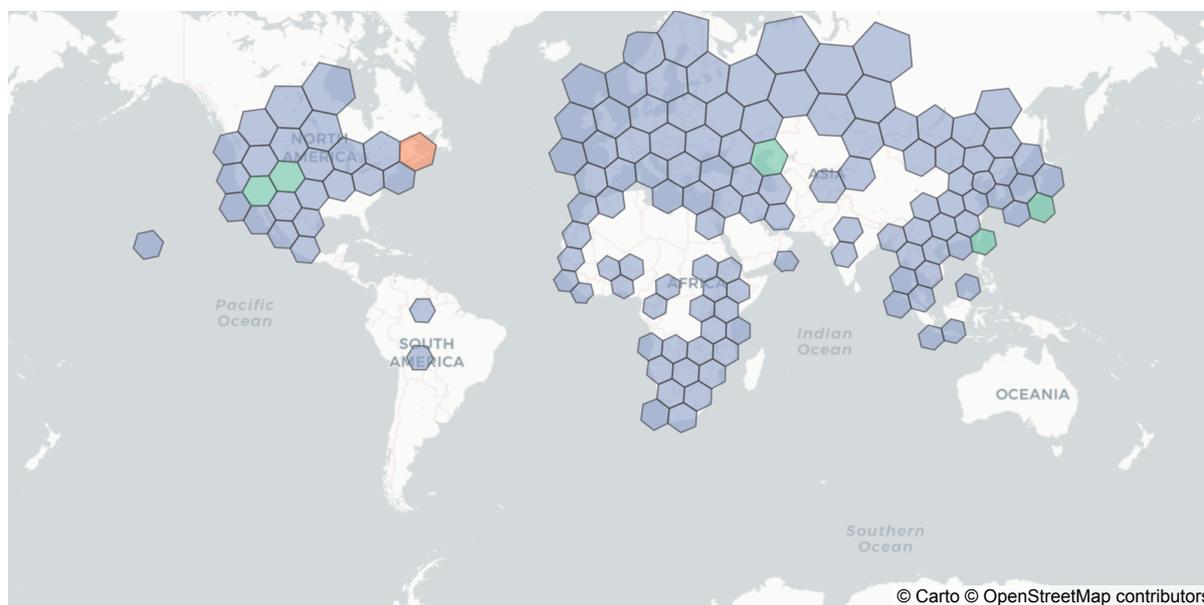
(b) Map

Figure S1: Distributions of anteater

The meaning of shapes and colours are explained in Figs. 2 and 3.



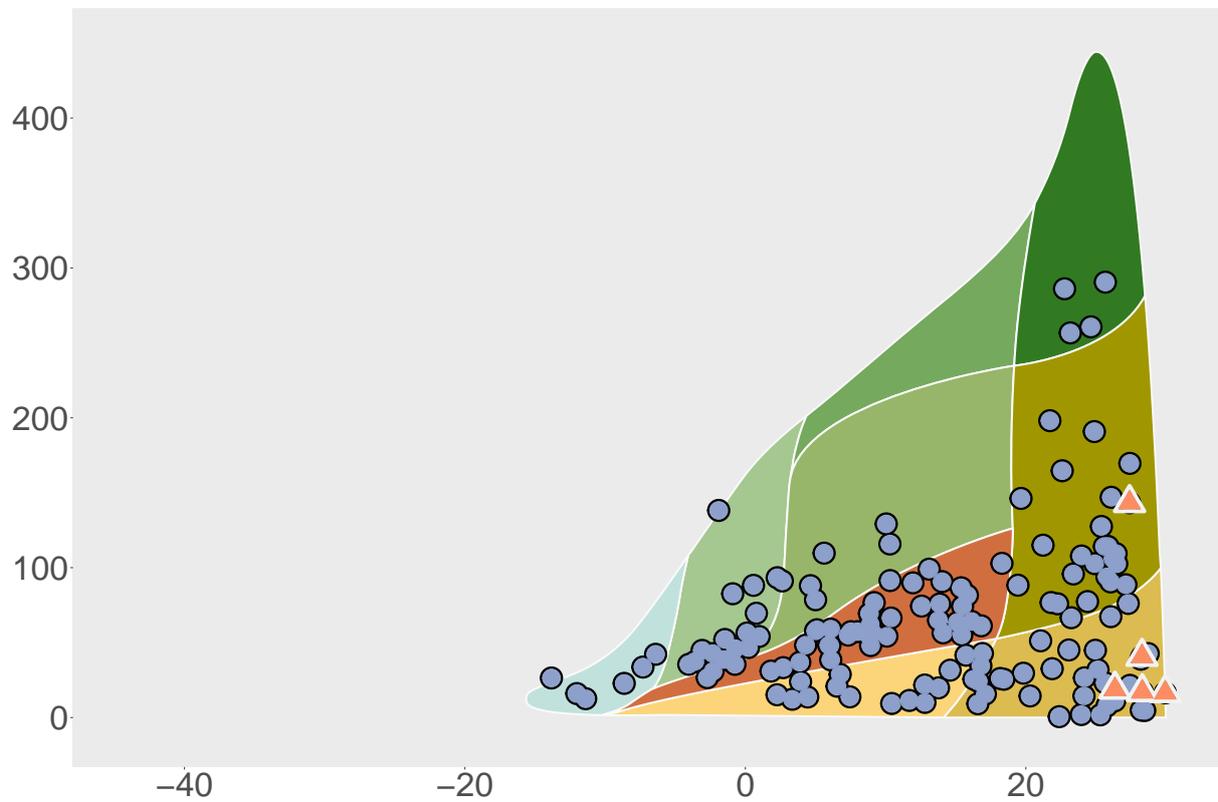
(a) Biome



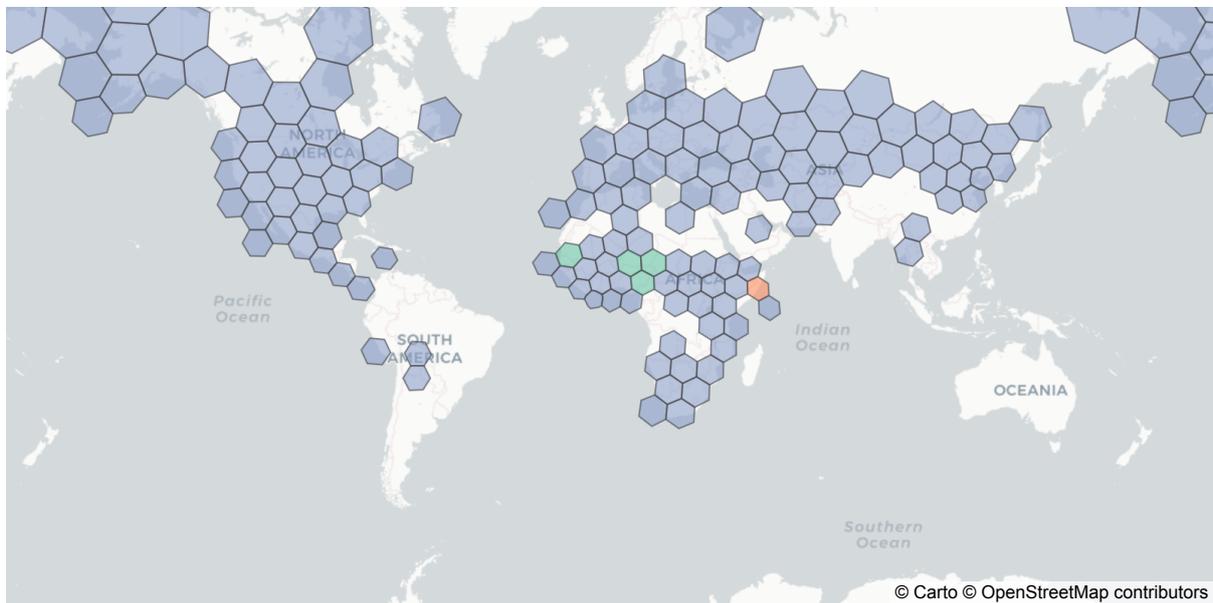
(b) Map

Figure S2: Distributions of badger

The meaning of shapes and colours are explained in Figs. 2 and 3.



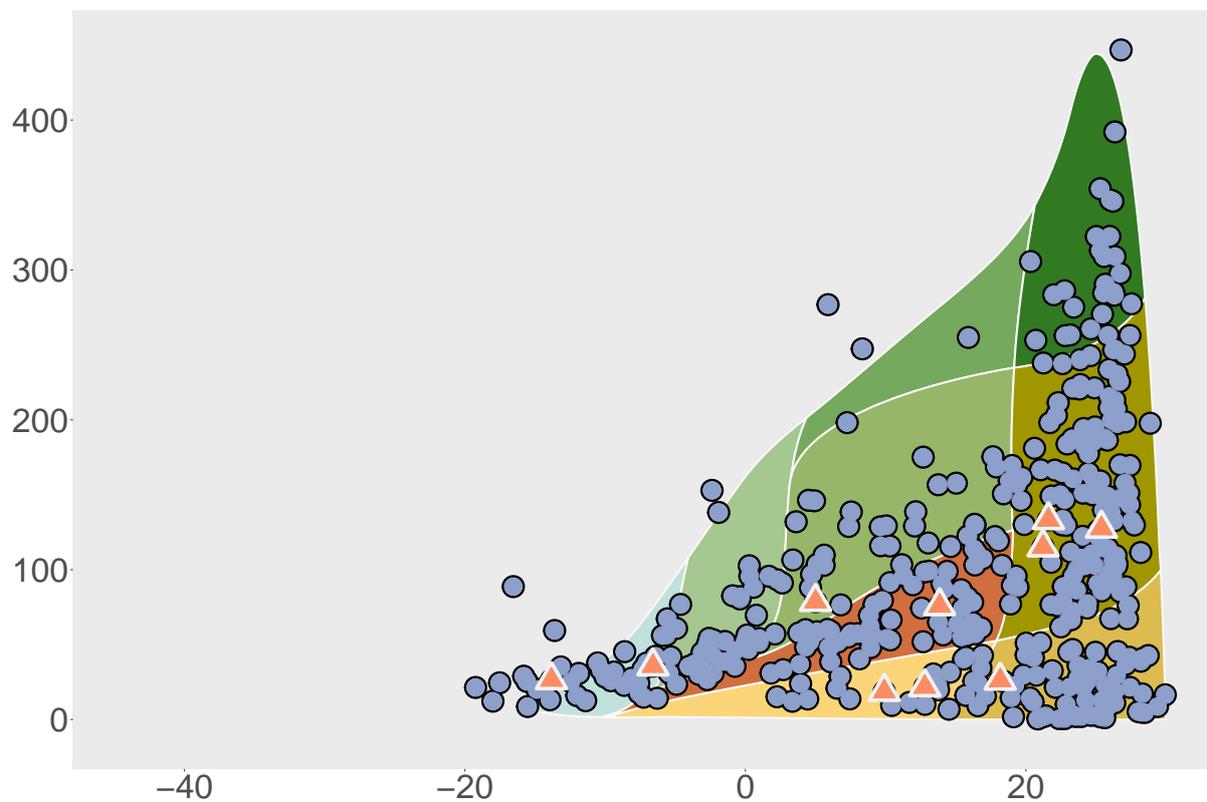
(a) Biome



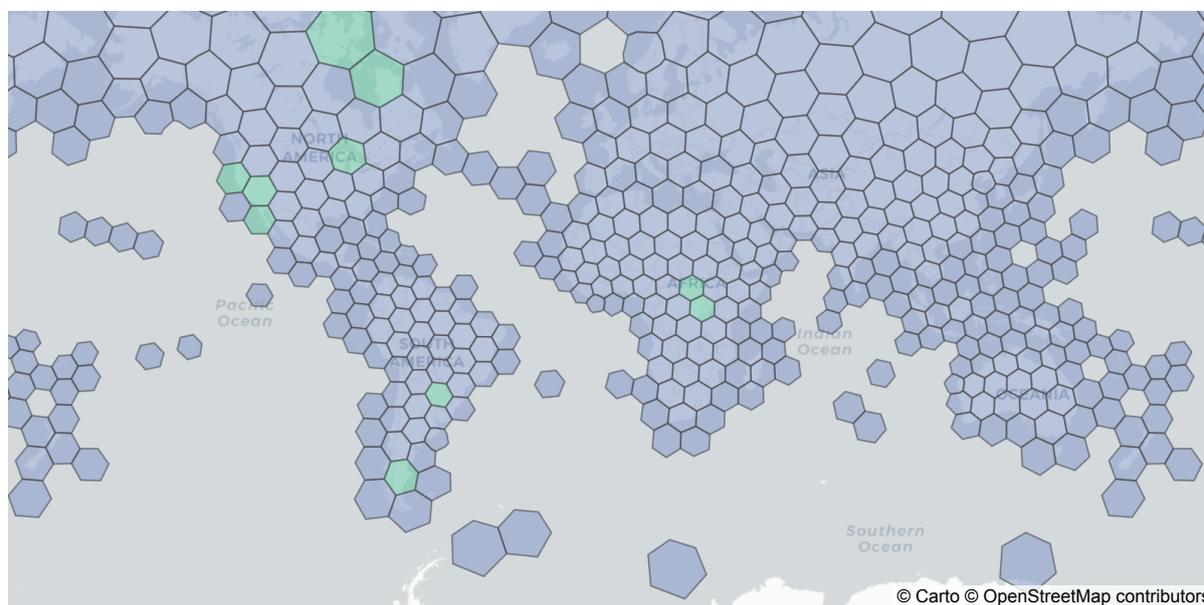
(b) Map

Figure S3: Distributions of ground squirrel

The meaning of shapes and colours are explained in Figs. 2 and 3.



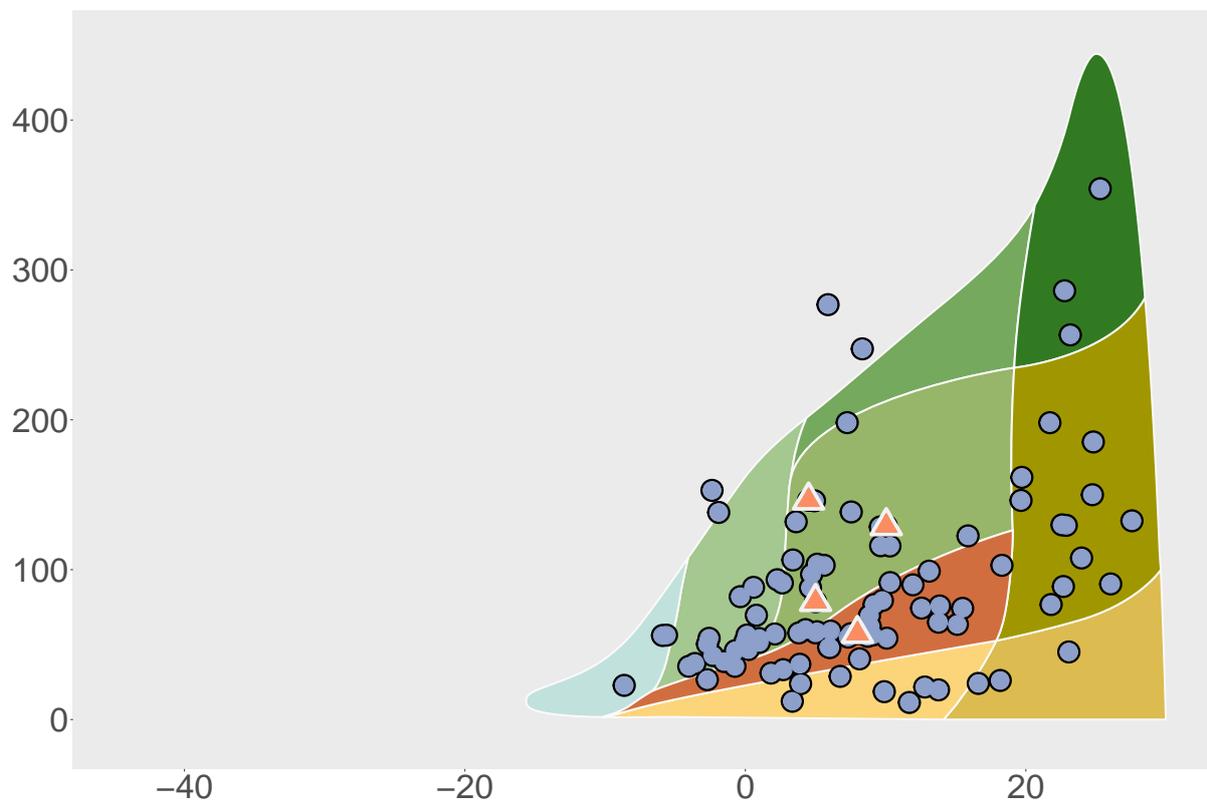
(a) Biome



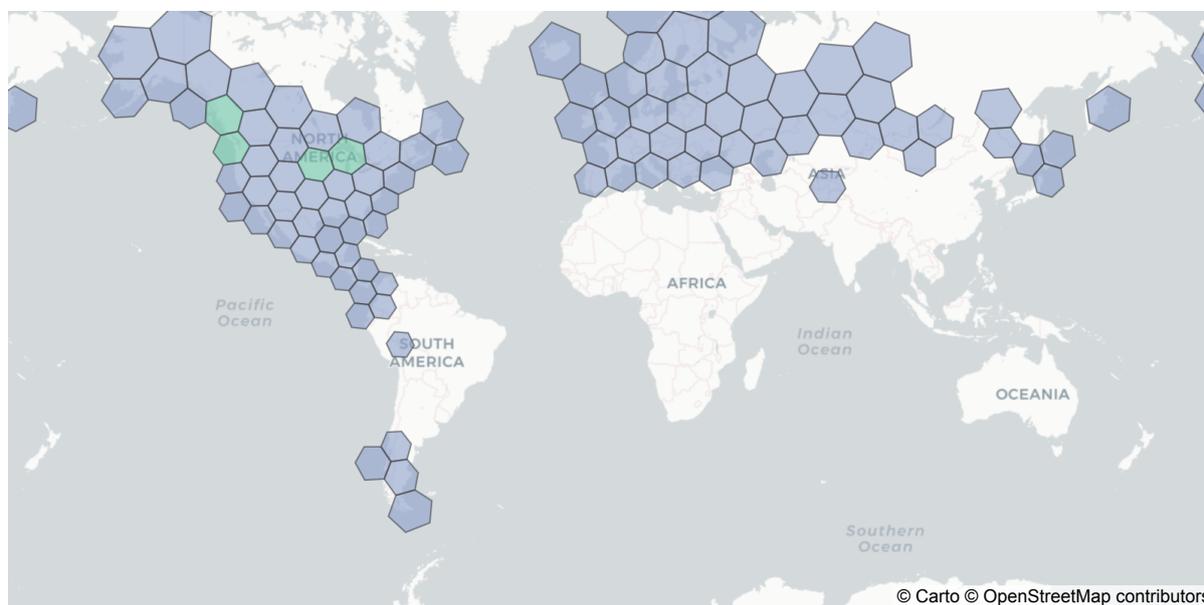
(b) Map

Figure S4: Distributions of hawk

The meaning of shapes and colours are explained in Figs. 2 and 3.



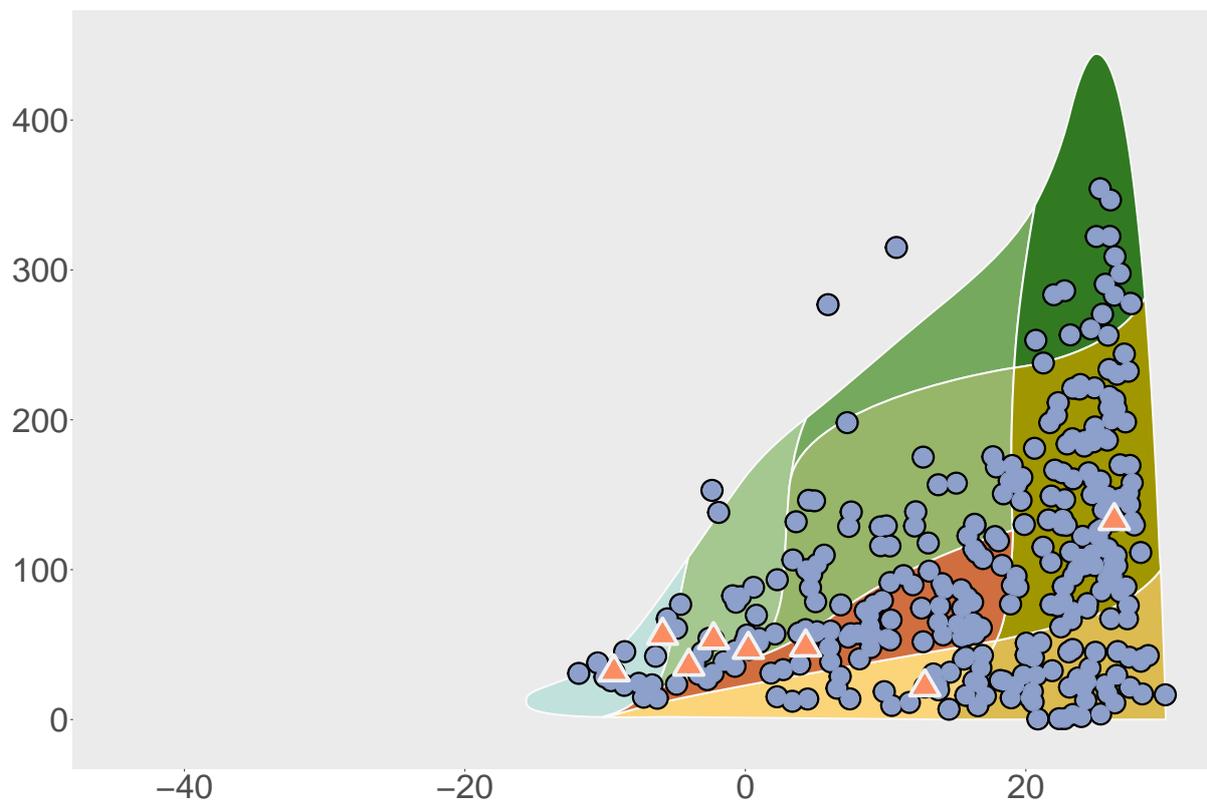
(a) Biome



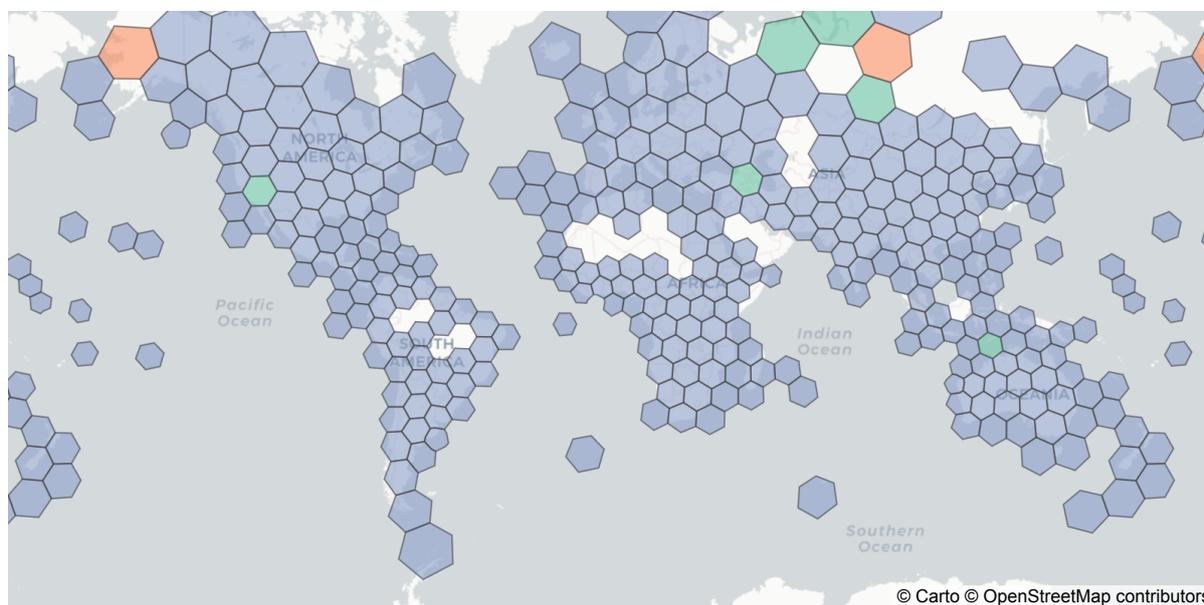
(b) Map

Figure S5: Distributions of mink

The meaning of shapes and colours are explained in Figs. 2 and 3.



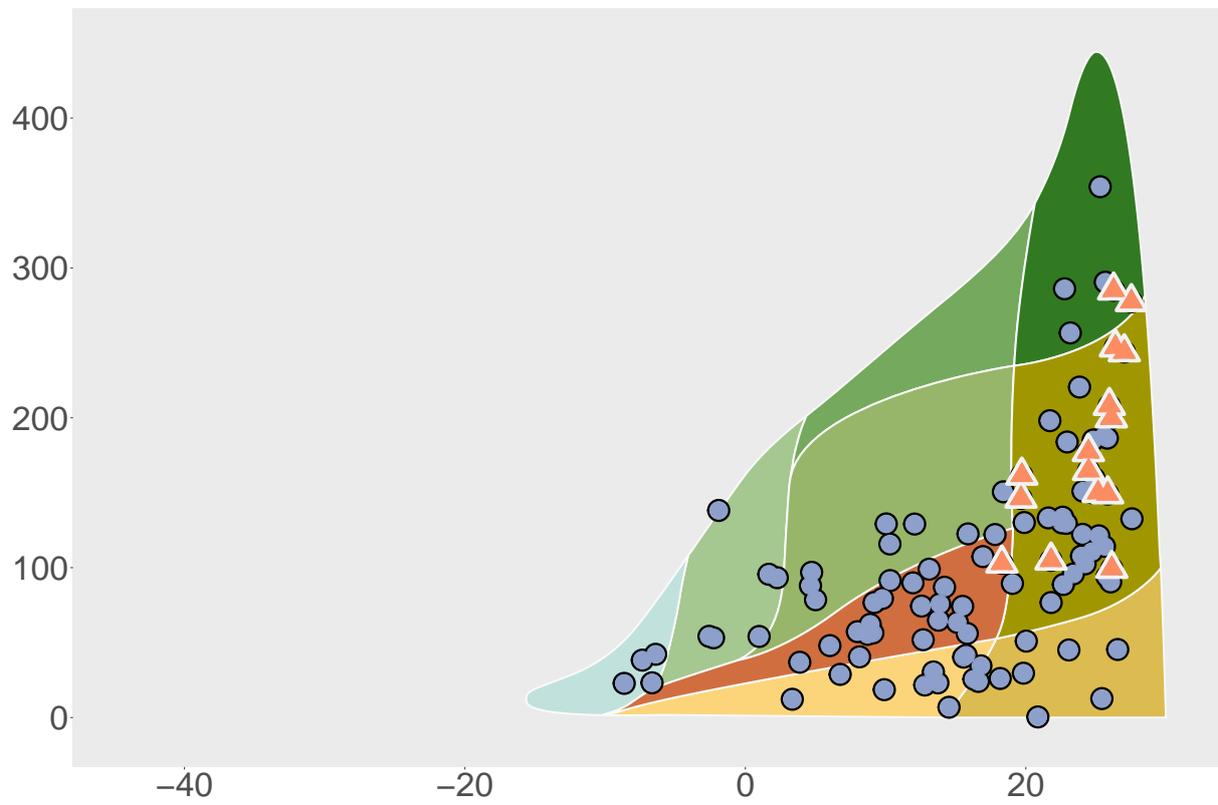
(a) Biome



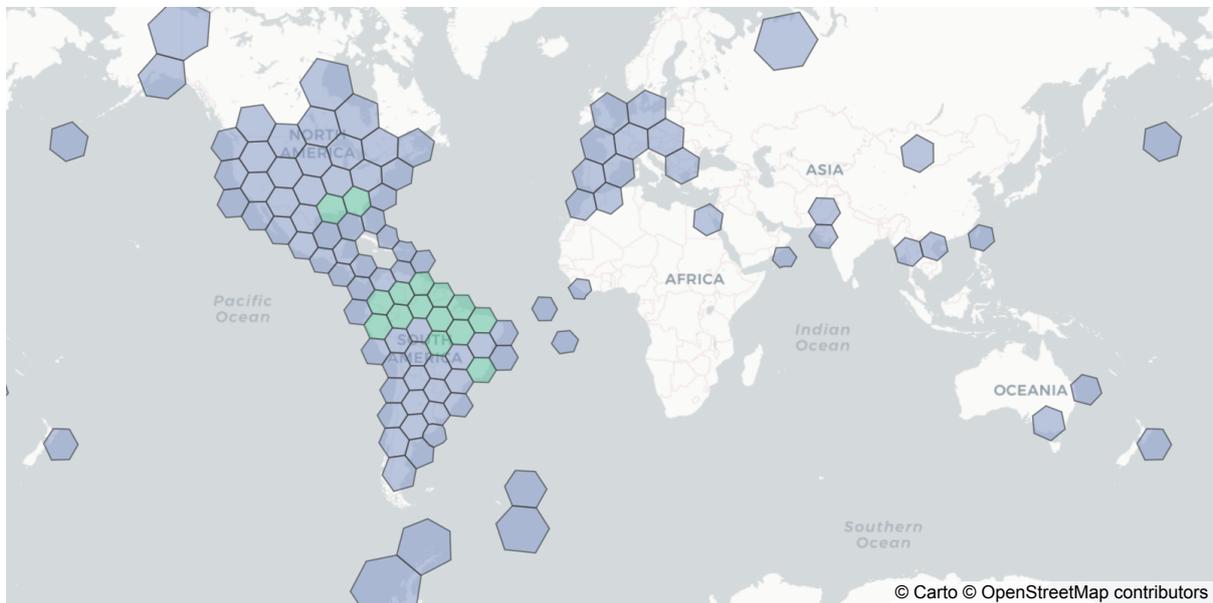
(b) Map

Figure S6: Distributions of mouse

The meaning of shapes and colours are explained in Figs. 2 and 3.



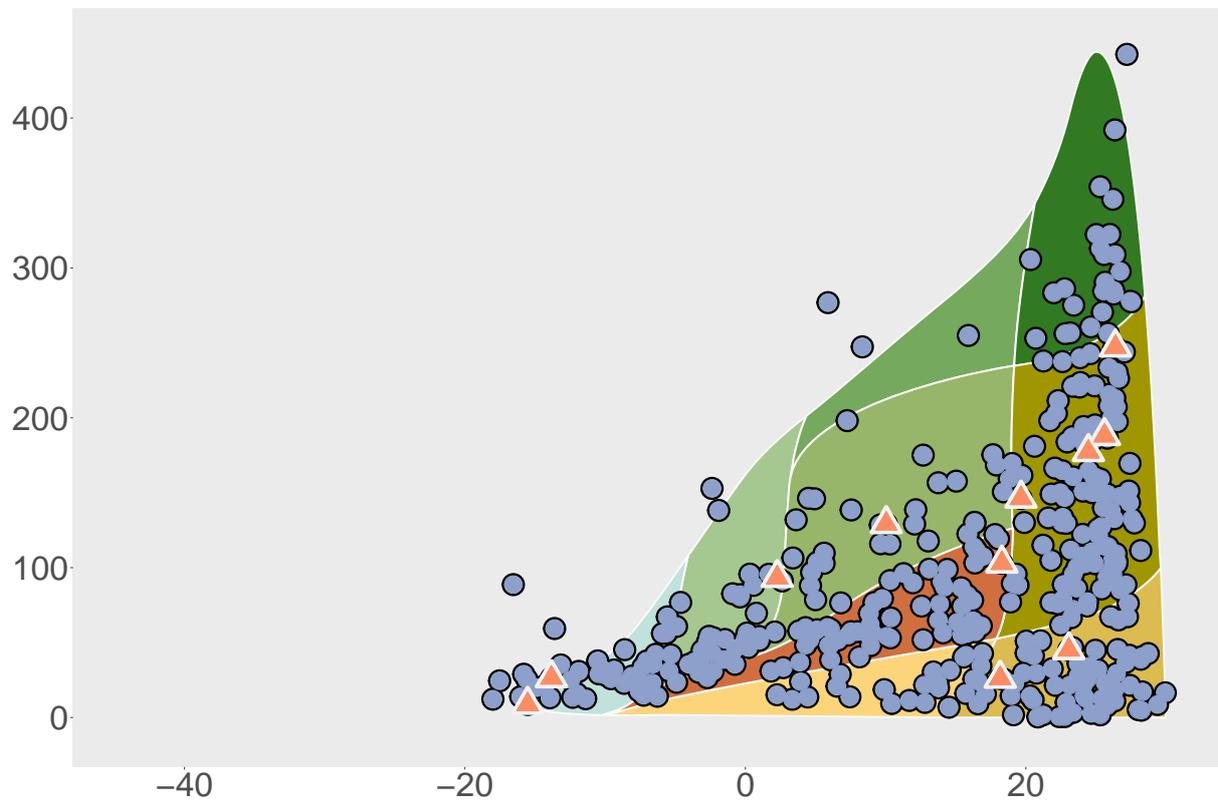
(a) Biome



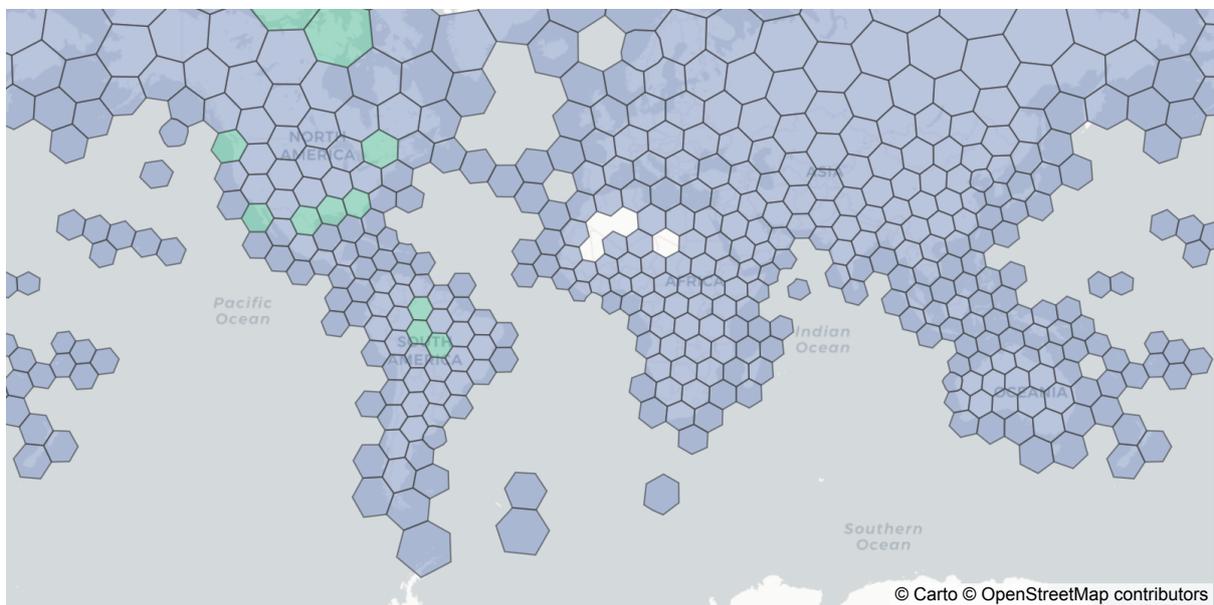
(b) Map

Figure S7: Distributions of opossum

The meaning of shapes and colours are explained in Figs. 2 and 3.



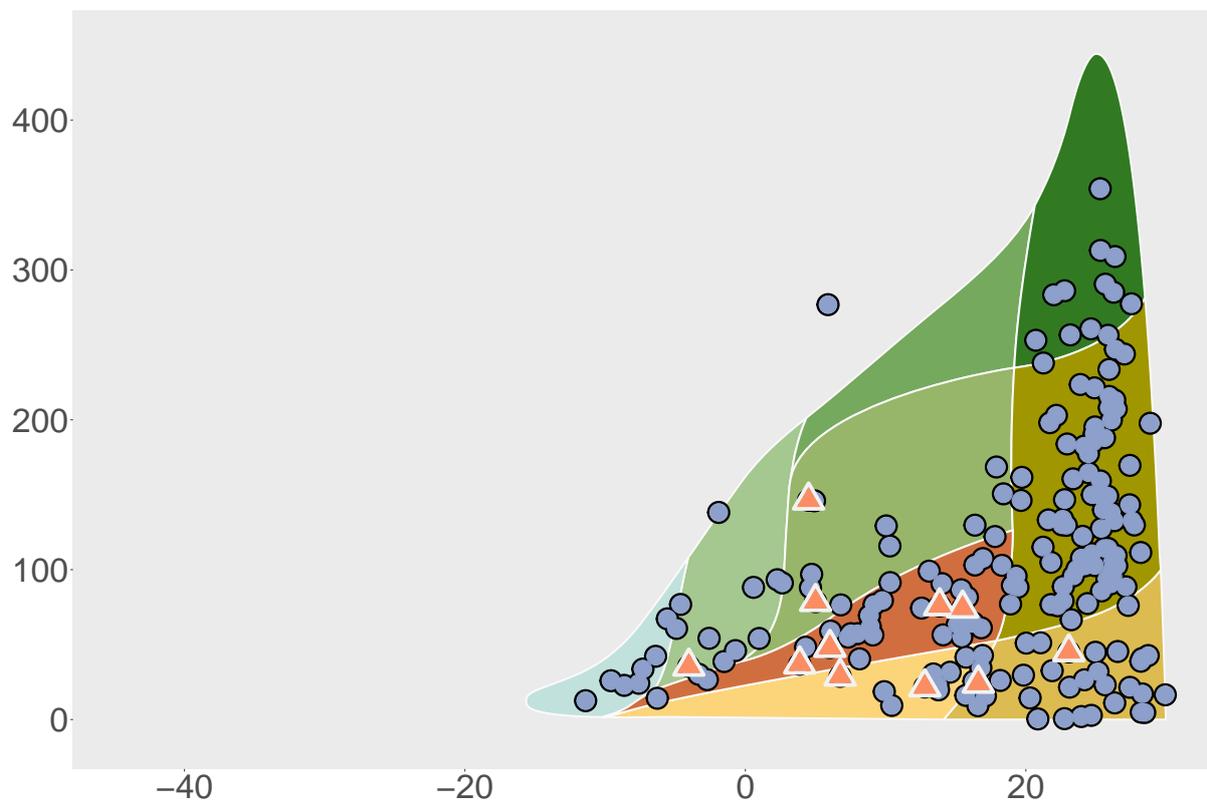
(a) Biome



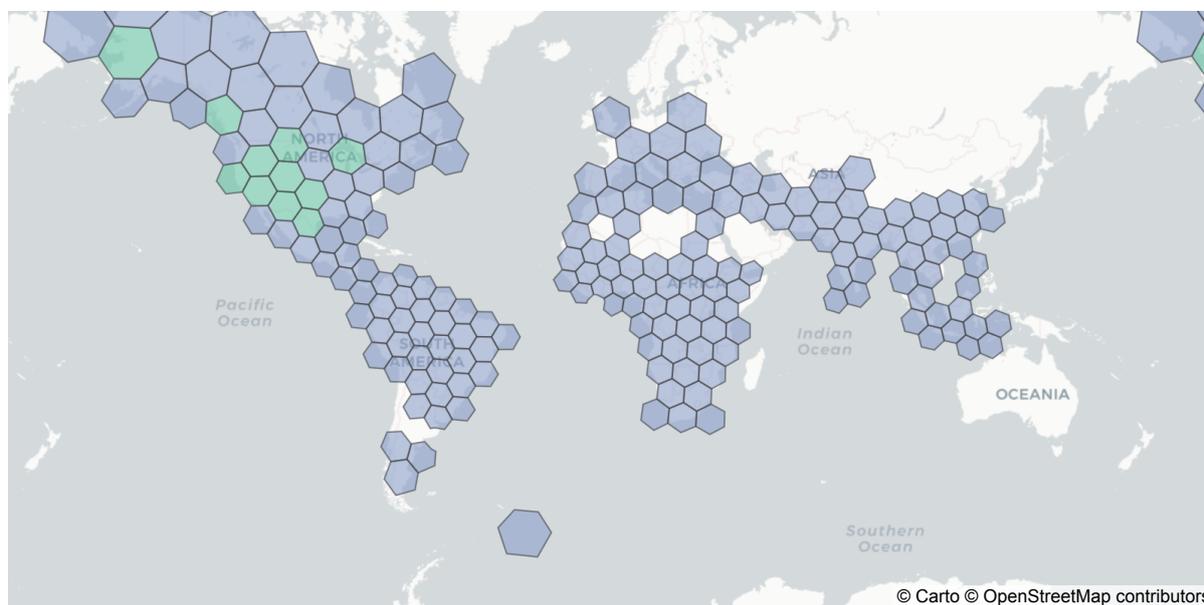
(b) Map

Figure S8: Distributions of owl

The meaning of shapes and colours are explained in Figs. 2 and 3.



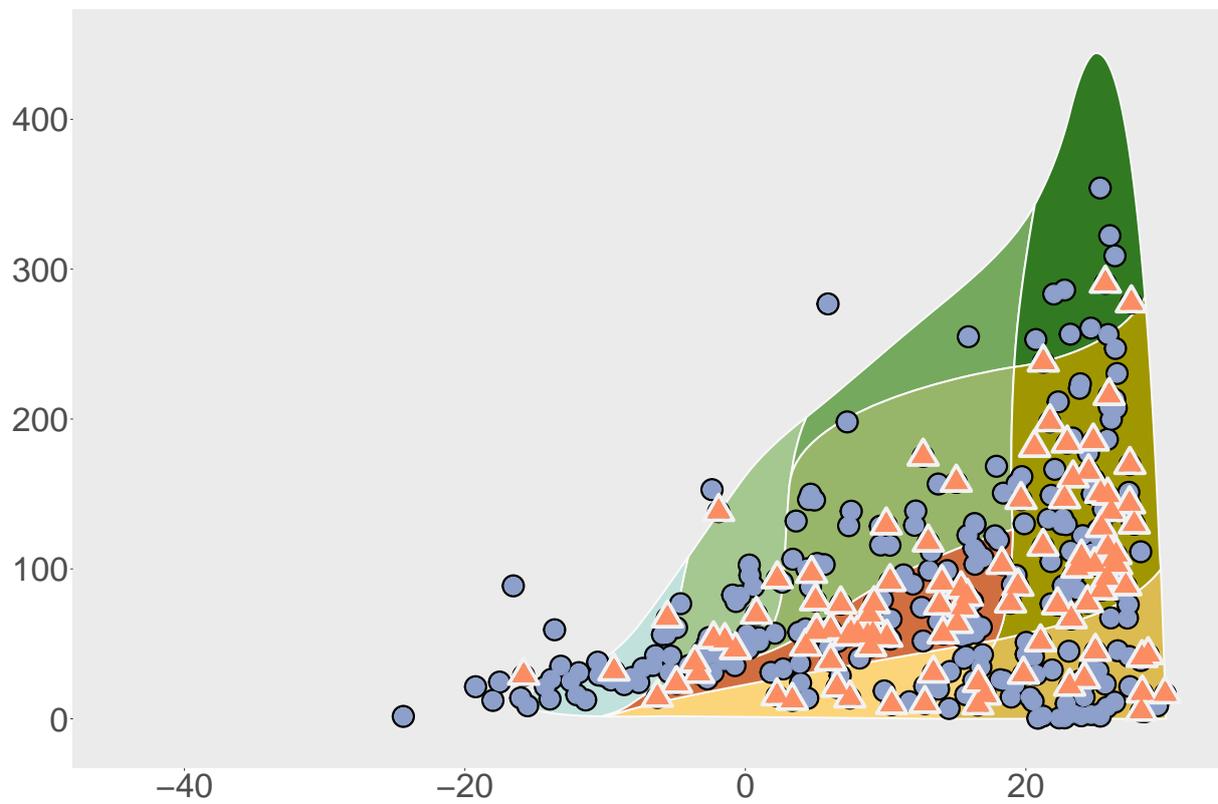
(a) Biome



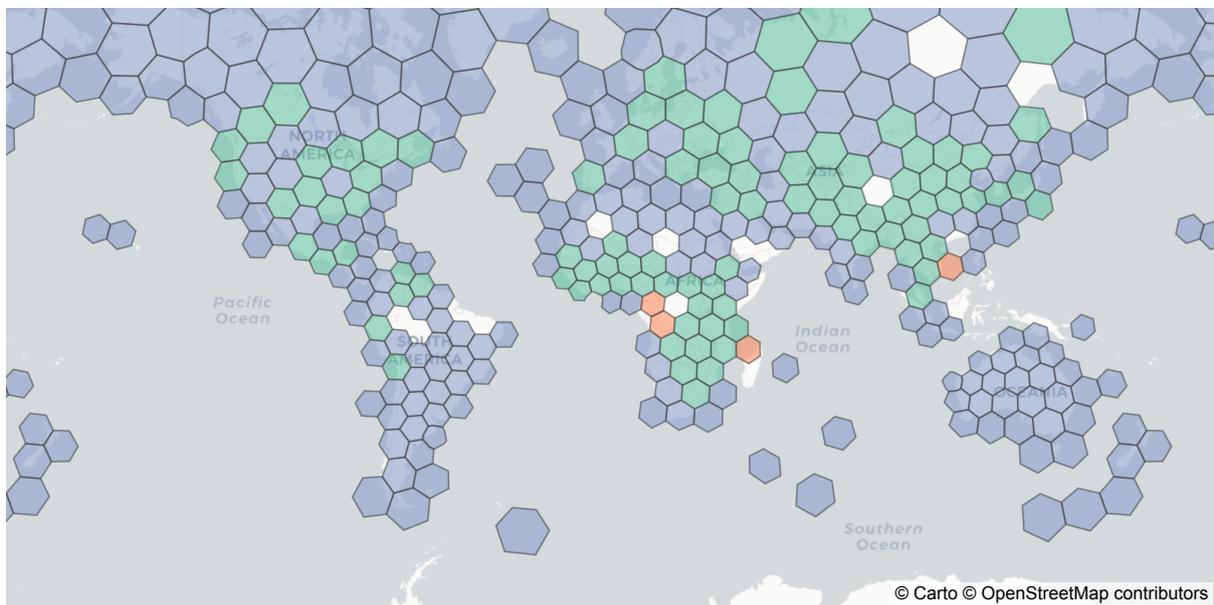
(b) Map

Figure S9: Distributions of porcupine

The meaning of shapes and colours are explained in Figs. 2 and 3.



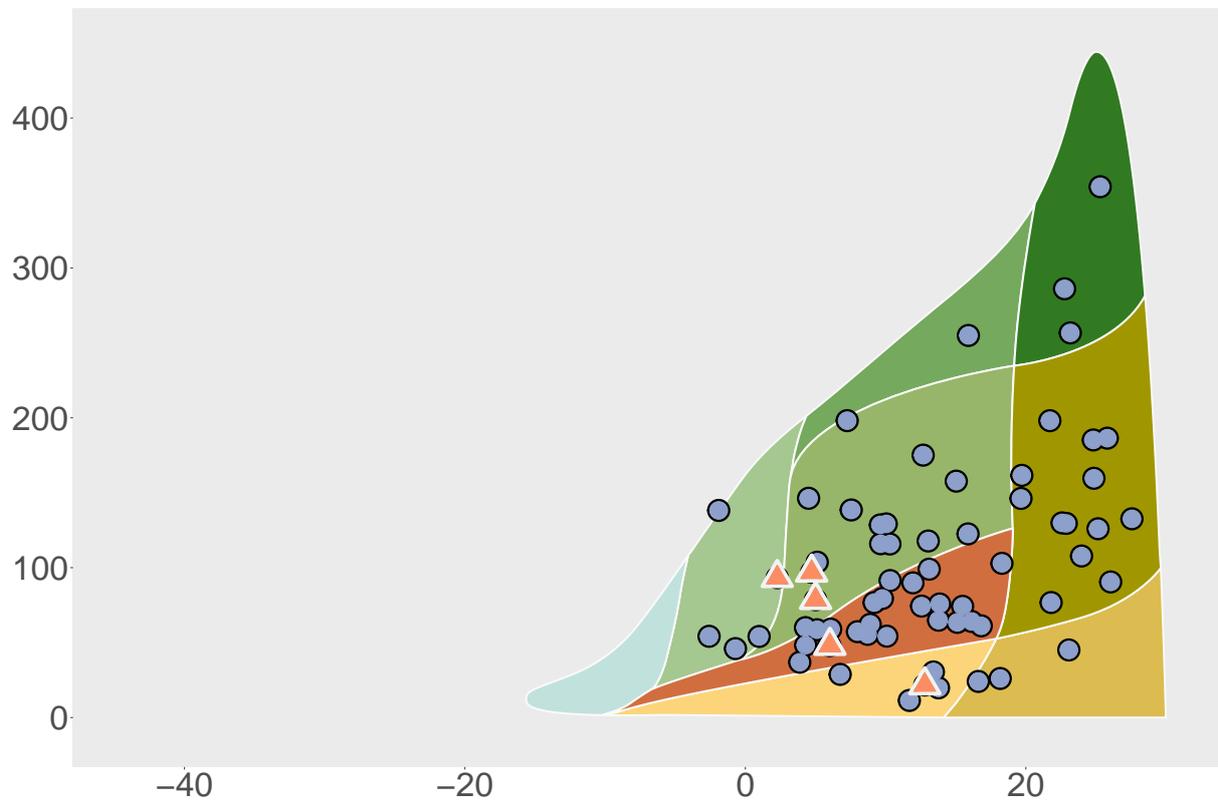
(a) Biome



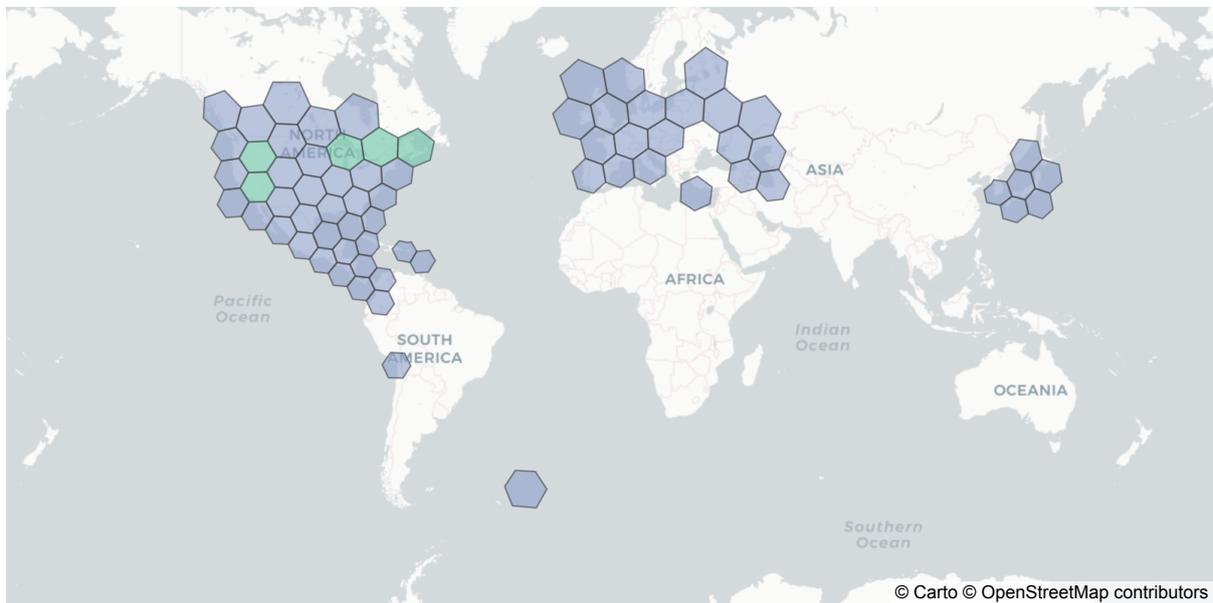
(b) Map

Figure S10: Distributions of rabbit/hare

The meaning of shapes and colours are explained in Figs. 2 and 3.



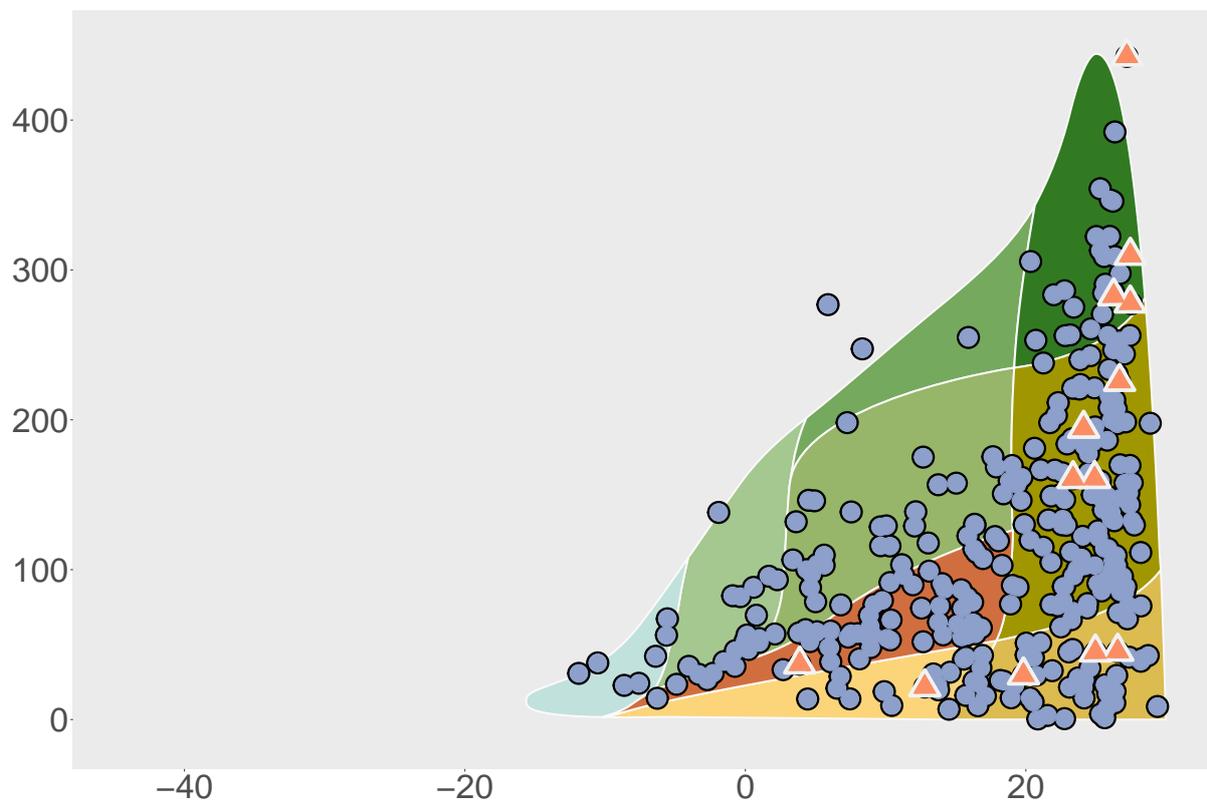
(a) Biome



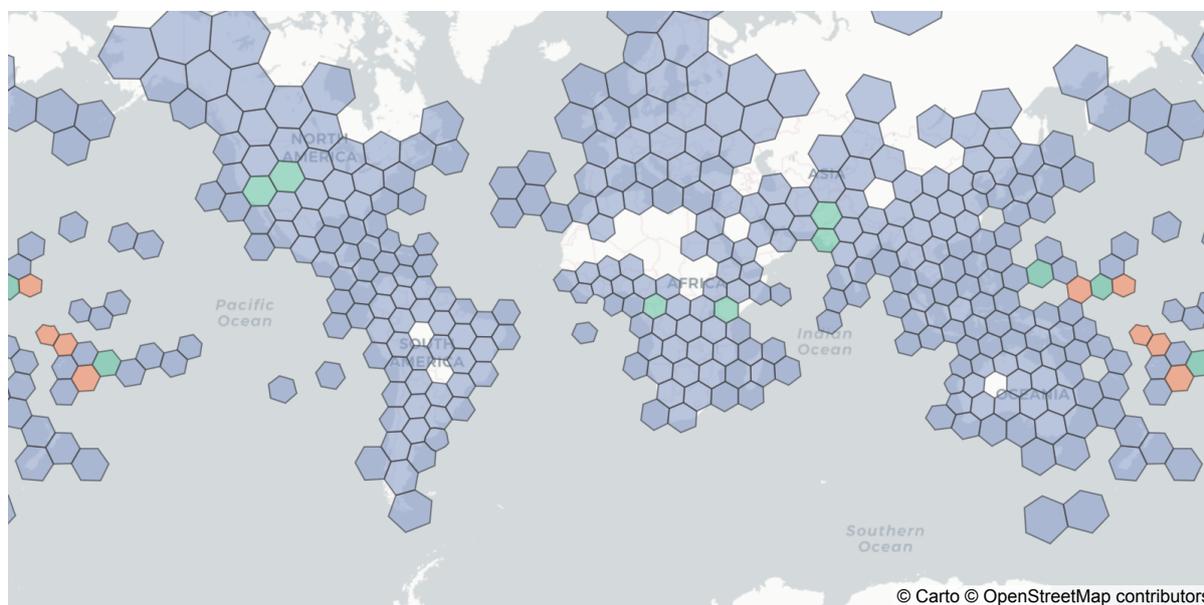
(b) Map

Figure S11: Distributions of racoon

The meaning of shapes and colours are explained in Figs. 2 and 3.



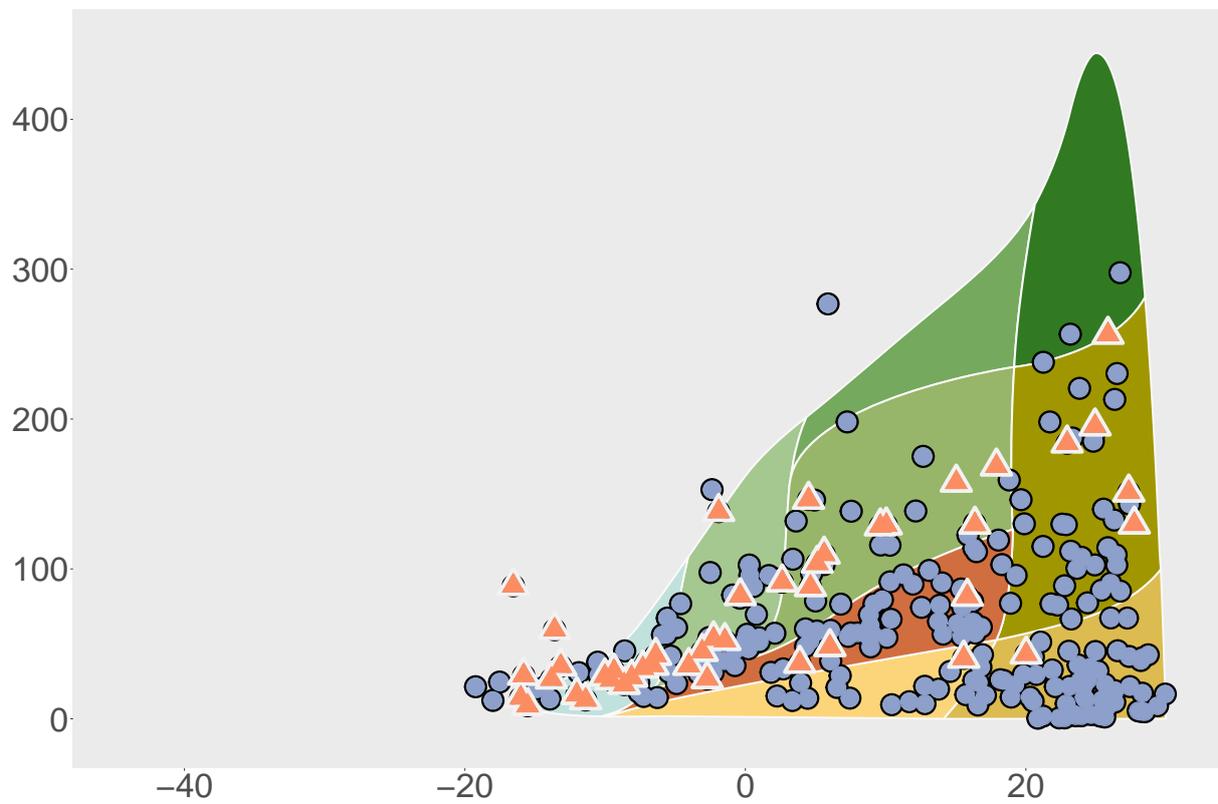
(a) Biome



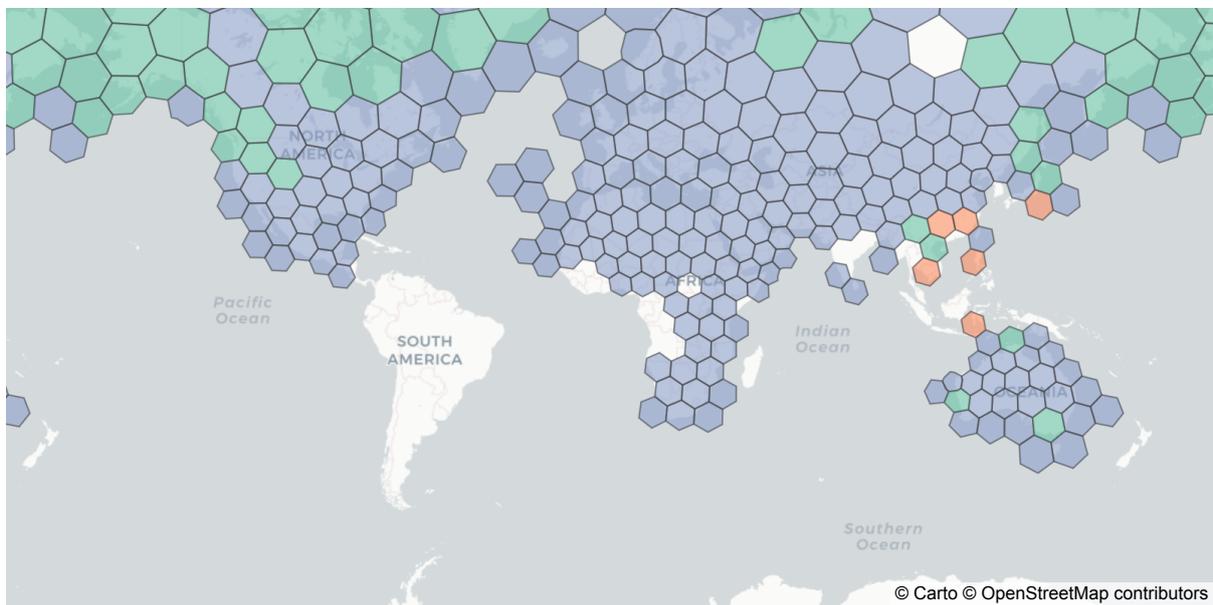
(b) Map

Figure S12: Distributions of rat

The meaning of shapes and colours are explained in Figs. 2 and 3.



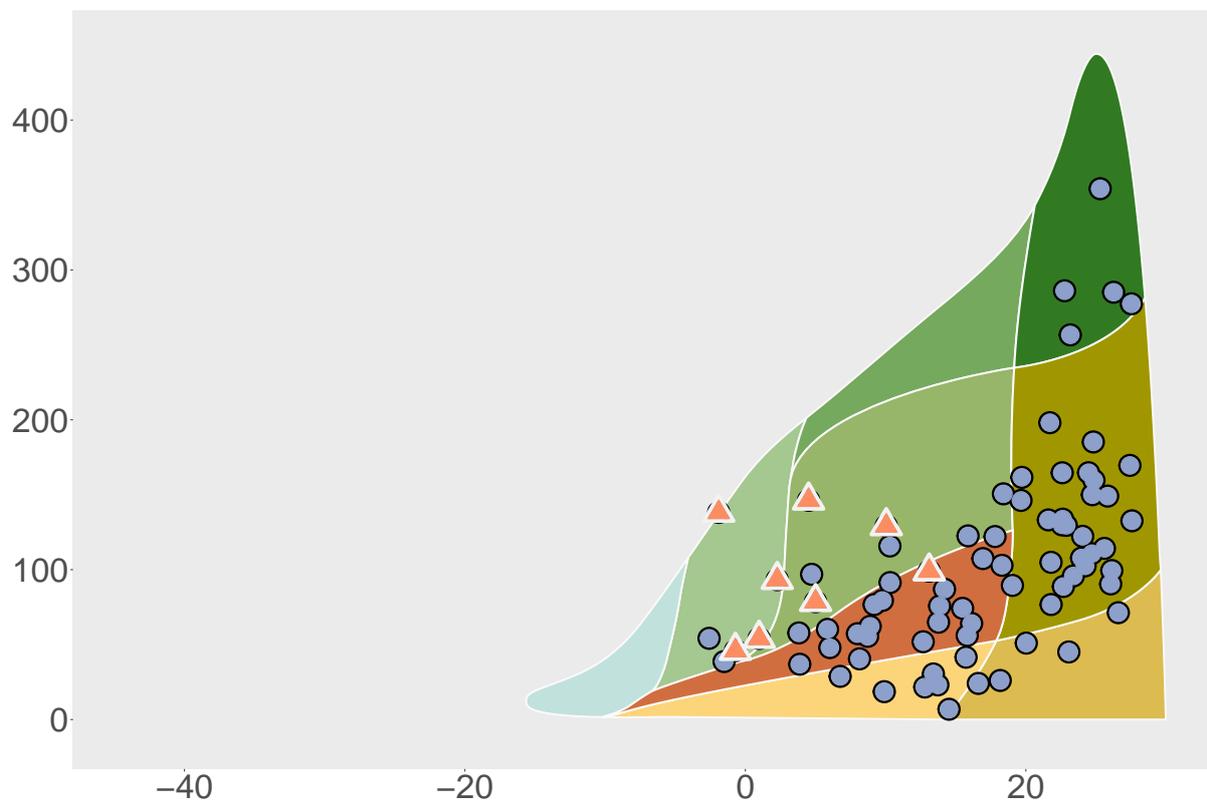
(a) Biome



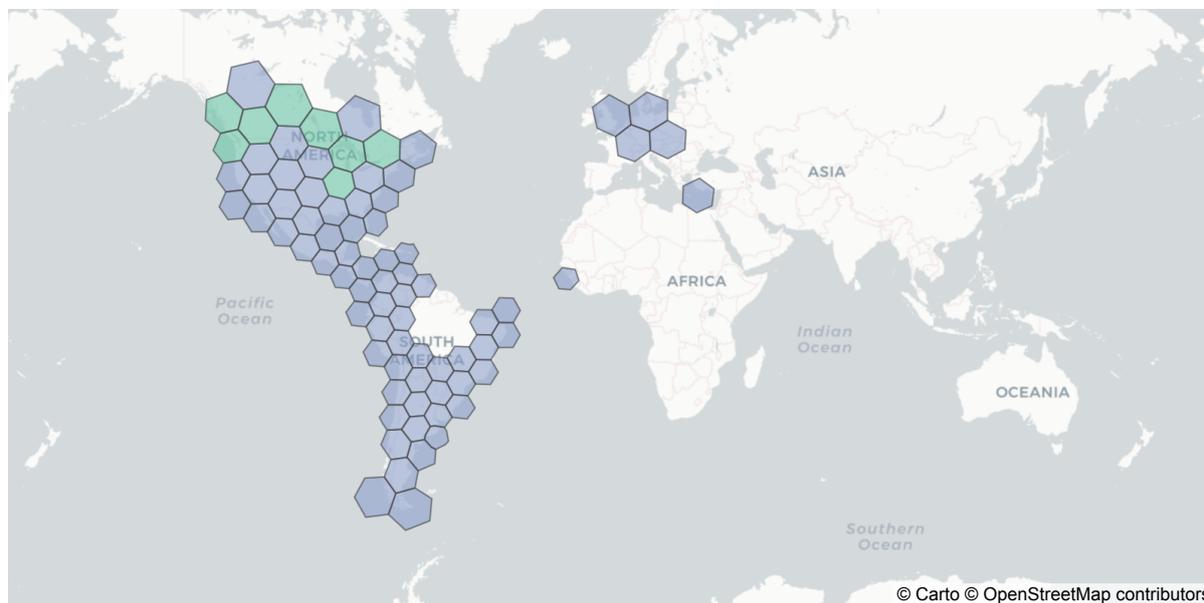
(b) Map

Figure S13: Distributions of raven/crow

The meaning of shapes and colours are explained in Figs. 2 and 3.



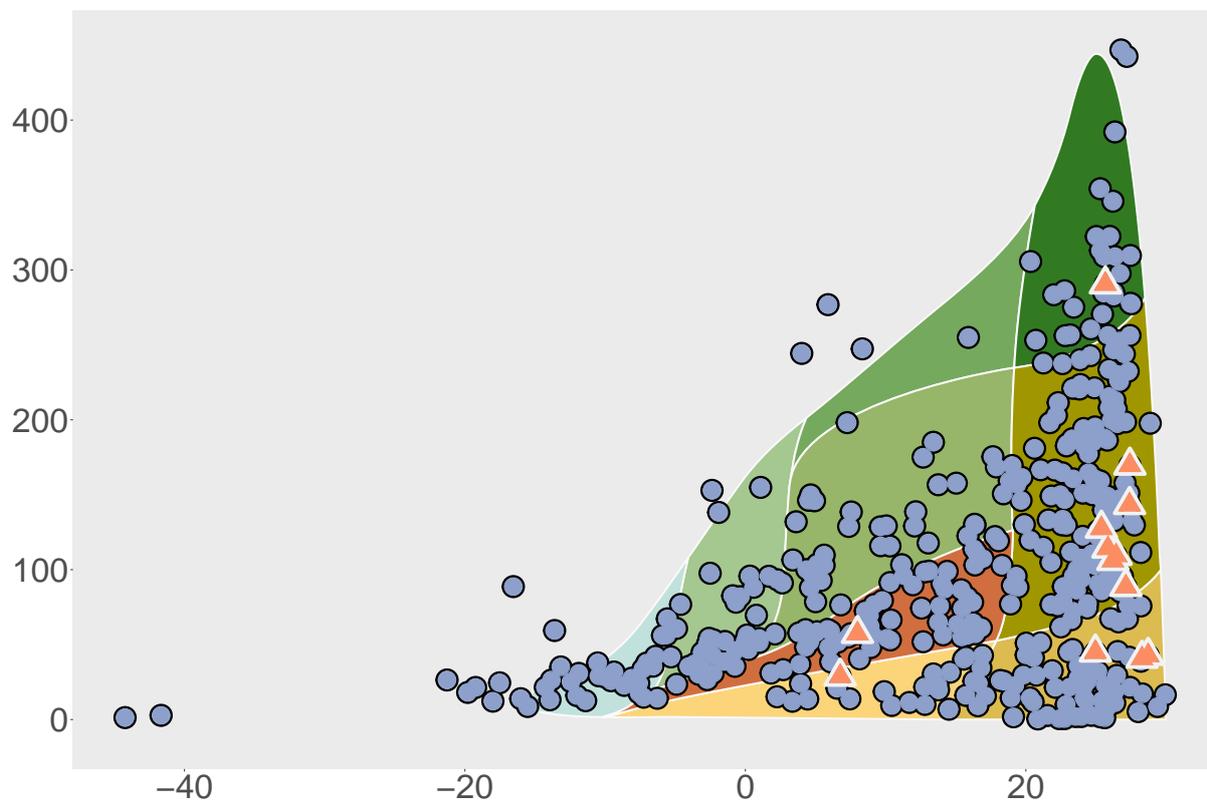
(a) Biome



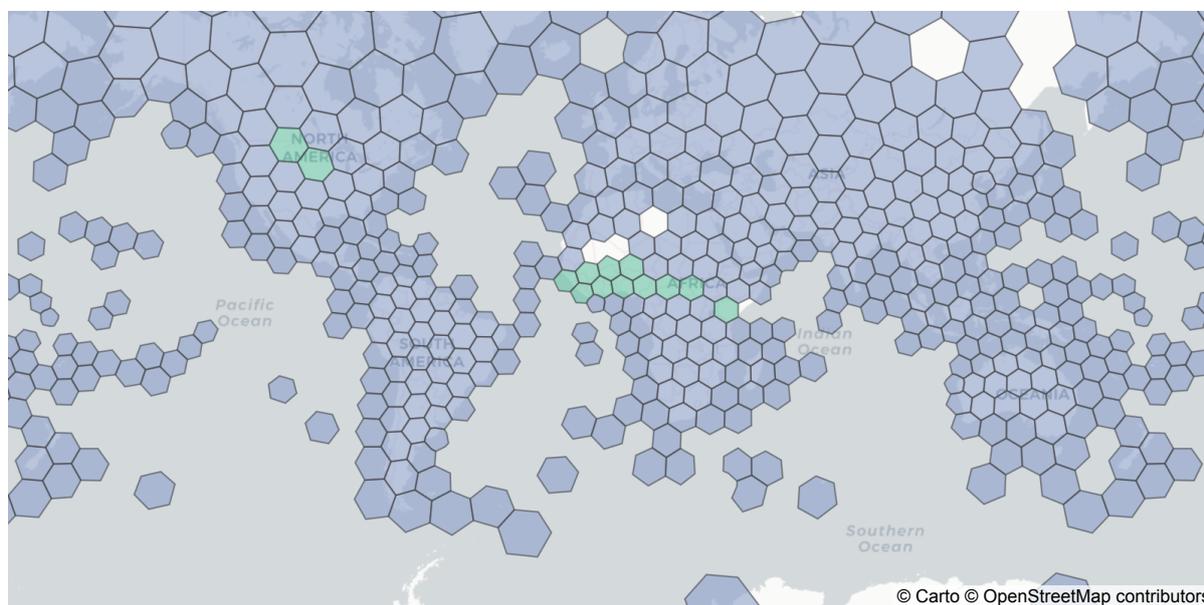
(b) Map

Figure S14: Distributions of skunk

The meaning of shapes and colours are explained in Figs. 2 and 3.



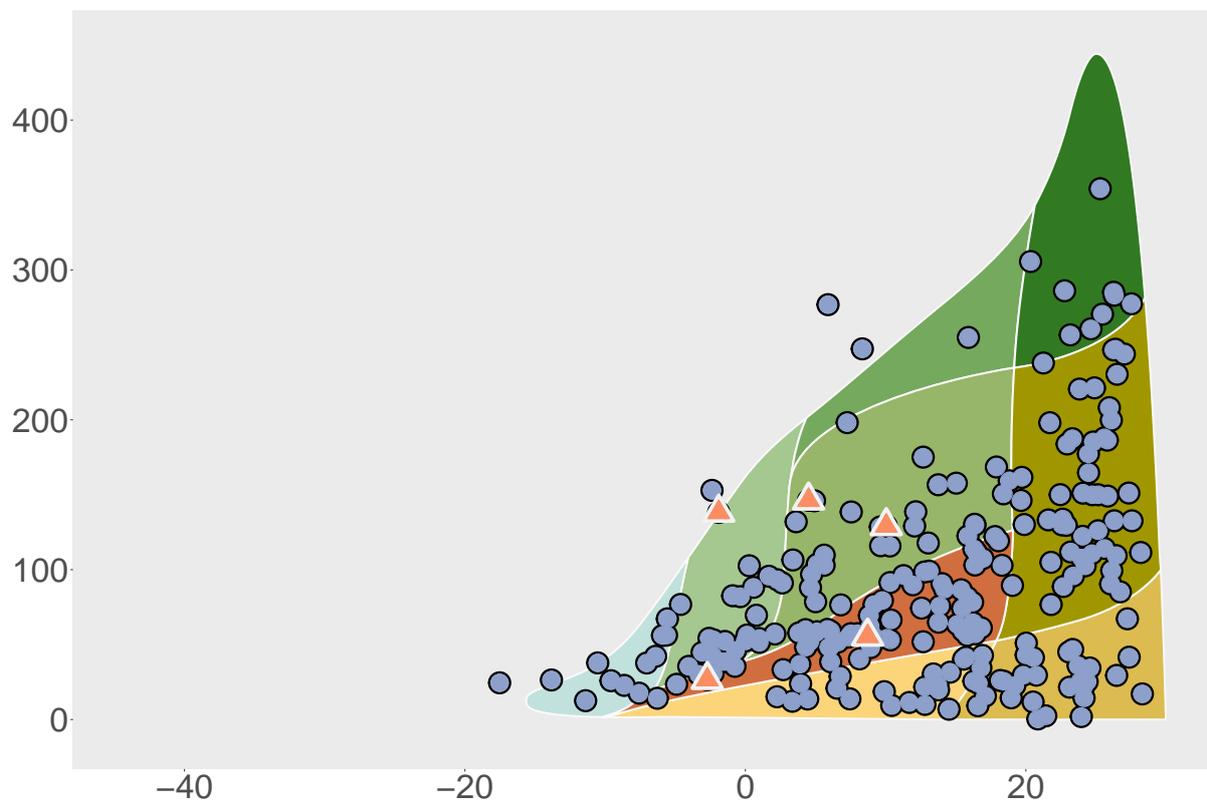
(a) Biome



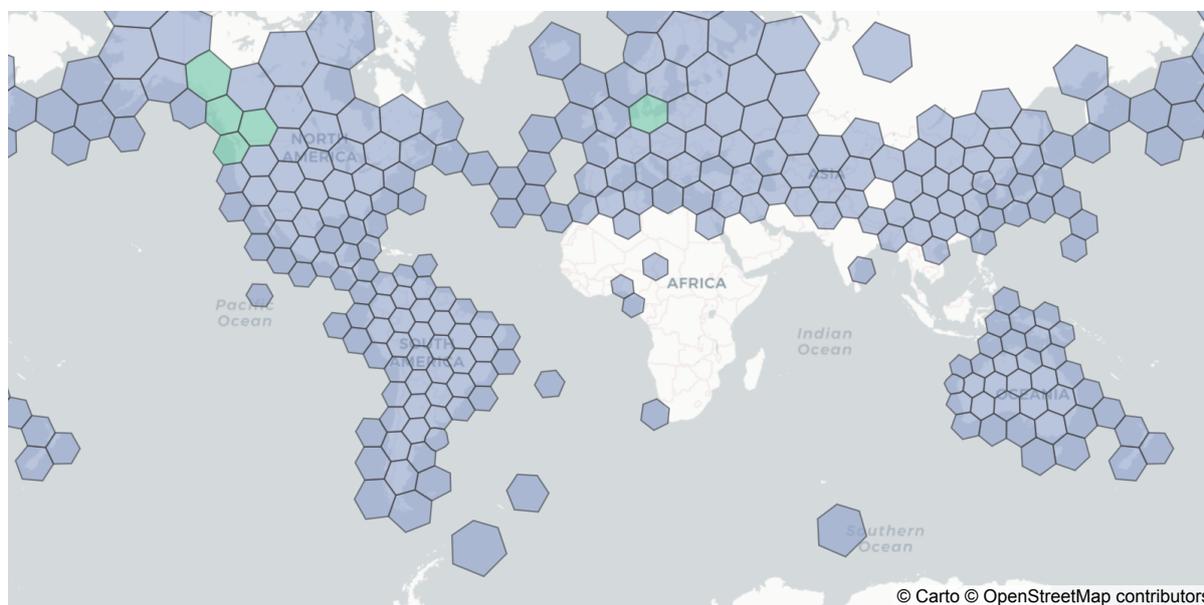
(b) Map

Figure S15: Distributions of spider

The meaning of shapes and colours are explained in Figs. 2 and 3.



(a) Biome



(b) Map

Figure S16: Distributions of wren

The meaning of shapes and colours are explained in Figs. 2 and 3.