Decomposing the Causes for Niche Differentiation Between Species Using Hypervolumes

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Hutchinson’s n-dimensional hypervolume concept holds a central role across different fields of ecology and evolution. The question of the amount of hypervolume overlap and differentiation between species is of great interest to understand the processes that drive niche dynamics, competitive interactions and, ultimately, community assembly. A framework is proposed to decompose overall differentiation among hypervolumes into two distinct components: niche shifts and niche contraction/expansion processes. Niche shift corresponds to the replacement of space between the hypervolumes occupied by two species, whereas niche contraction/expansion processes correspond to net differences between the amount of space enclosed by each hypervolume. A procedure to implement non-continuous trait data in the estimation of n-dimensional hypervolumes is proposed. Hypervolumes were constructed for three Darwin’s finches, Geospiza conirostris, Geospiza magnirostris, and Geospiza difficilis using intraspecific trait data. Results showed that significant niche shifts, not niche contraction, occurred between G. conirostris and G. magnirostris in Genovesa island, where they live in sympatry. This means that G. conirostris occupied a different niche space and not a reduced space on Genovesa. G. difficilis was well differentiated from the other two species. The proposed framework allows disentangling different processes underlying niche partitioning between coexisting species. This framework offers novel insights to understand the drivers of niche partitioning strategies among coexisting species.

Keywords: fundamental niche, hypervolume, morphospace, niche contraction, niche shift, realized niche

INTRODUCTION

The concept of species fundamental niche proposed by Hutchinson (1957) is at the roots of many ecological and evolutionary theories (e.g., Holt, 2009; Blonder, 2018). Hutchinson formalized this concept as a multidimensional hypervolume, defined by a set of n independent variables that represent biologically relevant axes. According to his perspective, each point within this n-dimensional space corresponds to a possible state of the environment permitting the species to survive.

Hutchinsonian niches can be estimated by quantifying the functional trait hypervolume occupied by the individuals of a given species. In essence, a set of functional traits, representing major axes of ecological strategies, are selected. Then, the hypervolume is constructed to estimate the morphospace occupied by the species. The hypervolume, representing the intraspecific trait variability, is commonly interpreted as the realized niche of the species (e.g., Pigot et al., 2016).
The question of the amount of morphospace overlap and differentiation between species is of great interest to understand the processes that drive niche dynamics, competitive interactions and, ultimately, community assembly (e.g., Ricklefs and Cox, 1977; Stubs and Wilson, 2004; Kraft et al., 2008). The hypervolume overlap corresponds to the shared space between two species, whilst hypervolume differentiation corresponds to the sum of the unique fractions of space belonging to each species. Niche differentiation between species may be caused by two distinct processes: niche shifts and niche contraction (or expansion). Niche shifts are determined by the replacement of niche space between species. In this case, species change their niche position in order to occupy a different niche space and reduce overlap. For example, changes of morphological traits related to food acquisition have been interpreted as a strategy to reduce niche overlap and avoid competition between species (Huey et al., 1974). On the other hand, niche contraction (or expansion) processes are determined by differences in niche breadth (e.g., Pulliam, 1986). Niche contraction occurs when a species reduces its niche breadth. For example, when faced with more intense competition from another species, many organisms restrict their utilization of shared microhabitats and/or other resources (Pianka, 2000). This might give them advantage exploring these resources over other, generalist, species, as they adapt to explore them to the fullest. Niche expansion occurs when a species augments its niche breadth when presented with ecological opportunity (McCormack and Smith, 2008). Compelling examples of niche expansion are given by studies in islands where species were released from competition (e.g., Lister, 1976). Thus, to understand the factors that drive niche differentiation between species it is important to disentangling niche shifts from niche expansion (or contraction) processes.

In this paper, we propose an integrated framework for partitioning hypervolumes via pairwise comparisons into niche shifts and niche contraction (or expansion). We also provide guidelines to incorporate non-continuous trait data into hypervolume estimation. We present a case study to illustrate how these concepts can be applied to decompose the causes for niche differentiation between species.

**MATERIALS AND METHODS**

**Hypervolume Partitioning**

To decompose hypervolume differentiation into different fractions, we use the set theory notation, in the manner of Hutchinson. Given a pairwise hypervolume system composed by hypervolumes $h_i$ and $h_j$, the total space occupied by them is given by their union ($h_i \cup h_j$). Total pairwise hypervolume corresponds to the sum of hypervolume overlap given by the interception ($h_i \cap h_j$) plus hypervolume differentiation, denoted by $h_i \Delta h_j$.

Thus, a pairwise hypervolume system can be partitioned according to the equation:

$$h_i \cup h_j = h_i \Delta h_j + h_i \cap h_j$$  \hspace{1cm} (1)

The differentiation between the hypervolumes $h_i$ and $h_j$ ($h_i \Delta h_j$) corresponds to the sum of their unique fractions:

$$h_i \Delta h_j = h_i \setminus h_j + h_j \setminus h_i$$ \hspace{1cm} (2)

This can also be obtained by:

$$h_i \Delta h_j = h_i \cup h_j - h_i \cap h_j.$$ \hspace{1cm} (3)

Total differentiation ($h_i \Delta h_j$) can occur as a consequence of two distinct processes: (i) the differentiation that results from the replacement of space between hypervolumes; and, (ii) differentiation that results from the net difference between the amounts of space enclosed by each hypervolume.

The net difference between both hypervolumes is given by $|h_i \setminus h_j|$. The replacement component can be obtained by subtracting the difference fraction from total differentiation, using Eq. 2:

$$h_i \setminus h_j + h_j \setminus h_i = |h_i \setminus h_j|$$ \hspace{1cm} (4)

This expression can be written in the form:

$$\max (h_i \setminus h_j, h_j \setminus h_i) + \min (h_i \setminus h_j, h_j \setminus h_i) - [\max (h_i \setminus h_j, h_j \setminus h_i) - \min (h_i \setminus h_j, h_j \setminus h_i)].$$ \hspace{1cm} (5)

By simplifying the expression, we obtain the replacement component: $2 \min(h_i \setminus h_j, h_j \setminus h_i)$.

Thus, the total differentiation component can be additively partitioned in two components:

$$h_i \Delta h_j = 2 \min(h_i \setminus h_j, h_j \setminus h_i) + |h_i \setminus h_j - h_j \setminus h_i|.$$ \hspace{1cm} (6)

A total partitioning of $h_i \cup h_j$ can be obtained by the equation (see Figure 1):

$$h_i \cup h_j = h_i \cap h_j + 2 \min(h_i \setminus h_j, h_j \setminus h_i) + |h_i \setminus h_j - h_j \setminus h_i|.$$ \hspace{1cm} (7)

The terms in Eqs 4 and 5 can be scaled in relation to the total space occupied by the hypervolume pairwise system, thus obtaining the following equivalency:

$$1 = (h_i \cap h_j + h_i \Delta h_j)/h_i \cup h_j = [h_i \cap h_j + 2 \min(h_i \setminus h_j, h_j \setminus h_i) + |h_i \setminus h_j - h_j \setminus h_i|]/h_i \cup h_j.$$ \hspace{1cm} (8)

This equivalency can be summarized in the equation:

$$1 = H_{overlap} + H_{diff} \text{total} = H_{overlap} + H_{repl} + H_{diff} \text{ repl}$$ \hspace{1cm} (9)

where, $H_{overlap}$ refers to the proportion of overlap between hypervolumes (similarity), $H_{diff} \text{ total}$ refers to total differentiation ($\delta$) between both hypervolumes, $H_{repl}$ corresponds to the fraction of differentiation that results from the replacement of space between hypervolumes and $H_{diff} \text{ repl}$ is the fraction of
Incorporating Different Data Types in Hypervolume Estimation

A limitation to the current hypervolume algorithms, is that they require datasets with continuous variables (Blonder et al., 2014; Blonder, 2018). We suggest that this limitation may be overcome by applying already available statistical methods to the matrix of species × traits (S × T), prior to hypervolume estimation.

Multivariate techniques, specially developed to deal with mixed data, may be used to ordinate the S × T matrix (e.g., Hill and Smith, 1976; Kiers, 1994). These ordination techniques are similar to a principal component analysis, allowing to extract several orthogonal axes (components) representing the variation of the S × T matrix (e.g., PCAMIX, Chavent et al., 2017). These axes can be interpreted in relation to the original variables (traits), by examining the matrix of eigenvectors in a similar way to a PCA. Then, the resulting axes are used as the new variables to compute hypervolumes (Figure 2). If the aim is to include all the variation of the S × T matrix in the analyses, one should retain all the axes. However, most probably some of these components represent little variation and are difficult to interpret. Therefore, a more parsimonious strategy would be to select only those axes that provide a clear interpretable result.

The distance-based framework proposed by Laliberté and Legendre (2010) is an alternative to consider. Briefly, this approach is based on three steps: (i) an appropriate distance measure (Gower dissimilarity measure; Gower, 1971; Podani, 1999; Legendre and Legendre, 2012) is computed from the S × T matrix; (ii) this distance matrix is analyzed through principal coordinate analysis (PCoA); and, (3) the resulting PCoA axes are used as the new variables to compute hypervolumes (Figure 2). A comprehensive description of PCoA is given in Legendre and Legendre (2012). Nevertheless, PCoA may be difficult to implement in the context of hypervolume estimation, since it tends to produce n-1 eigenvectors (n = the number of observations). This means that for some ecological datasets, the calculation of hypervolumes based on all PCoA axes may be impractical or even impossible. Therefore, one may choose the first few axes of the PCoA, provided that they represent a large proportion of variation of the original data set. Another possibility is to use non-metric multidimensional scaling (NMDS), instead of PCoA, to analyze the Gower's distance matrix. In NMDS one could choose a small number of axes to represent the underlying dissimilarity between observations. The NMDS tries to find, interactively, the optimal configuration in the reduced space that maximizes the observation distances in the configuration and the original distances (for a comprehensive review see Legendre and Legendre, 2012). Frequently, a small number of axes is sufficient to represent the original distances.

All techniques allow to extract orthogonal axes, which eliminates the potential correlation among trait variables, a pre-requisite of hypervolume methods. Independently of which technique is used to transform the S × T matrix, an additional point to consider is giving weights to trait variables, because a single categorical trait can be coded by several dummy variables. Also, prior weights can be given to trait variables, allowing to differentiate traits that contribute more to the

Assessing Intraspecific Trait Variability and Niche Partitioning

Here, we advocate that intraspecific trait variability can be used to estimate niche parameters. The underlying assumption of a trait-based approach is that traits reflect species adaptations to the environment (Diaz and Cabido, 2001) and, hence, should be a useful tool to quantify species’ niches (Vielle and Jiang, 2009).

In other words, we are assuming that the morphospace occupied by each species is a surrogate for its realized niche.

Consider N individuals belonging to a given species s and measure a set of T traits for each individual. The matrix S × T can be used to construct the hypervolume, which is assumed to describe the realized niche of the species. Therefore, the niche breadth of species s is given by the extent of the hypervolume.

Obviously, hypervolumes can be calculated for a set of species (S). The set operations can be applied to hypervolumes using Eqs 1–7 for each species pair. This allows obtaining a pairwise partitioning of the niche species pairs. Therefore, the terms in Eq. 7 mean: $H_{\text{overlap}} =$ niche overlap between two species; $H_{\text{total}} =$ total niche differentiation between two species; $H_{\text{repl}} =$ differentiation due to the replacement of niche space; $H_{\text{diff}} =$ differentiation due to differences of niche breadth between species. Thus, higher values of $H_{\text{repl}}$ are indicative of niche shifts between species, whereas higher values of $H_{\text{diff}}$ are indicative of niche contraction/expansion of one of the species in relation to the other.

FIGURE 1 | Diagram showing hypervolume partitioning in two dimensions. The overlap between hypervolumes $h_i$ and $h_j$ is given by the interception ($h_i \cap h_j$) and the differentiation ($h_i \Delta h_j$) is given by the sum of their unique fractions, $h_i \setminus h_j$ and $h_j \setminus h_i$. Note that: $h_i \Delta h_j = 2 \min(h_i \setminus h_j, h_j \setminus h_i) + |h_i \setminus h_j| + |h_j \setminus h_i|$, where $2 \min(h_i \setminus h_j, h_j \setminus h_i)$ corresponds to the differentiation due to the replacement of space (gray shaded area) and $|h_i \setminus h_j| + |h_j \setminus h_i|$ refers to the net difference between both hypervolumes (white area in the $h_i \setminus h_j$ space).
biological performance of the species. Contrary to PCAMIX, the information of each trait is lost during the PCoA and NMDS, because the \( S \times T \) matrix is transformed into a dissimilarity matrix of individuals, prior to the calculation of hypervolumes. Thus, the interpretation of these shapes is relative to the axes of the PCoA and NMDS and not in relation to the traits. Therefore, questions about which traits contribute more or less to a given hypervolume parameter are difficult to address.

**Case Study**

We present a demonstration analysis of hypervolume partitioning for three species of Darwin’s finches: *Geospiza conirostris*, *Geospiza magnirostris* and *G. difficilis*. *G. conirostris* and *G. magnirostris* species inhabit Island Genovesa, but only the former is present in Island Española. A well-known hypothesis for the two species is that changes of morphological traits induced by interspecific competition occurred on Genovesa and, therefore, they should have evolved to occupy dissimilar niches (Lack, 1945, 1947; Grant and Grant, 1982). Therefore, we hypothesize that niche differentiation between these species was determined by the replacement of space (niche shift) and not the contraction/expansion of niches. *G. difficilis* inhabits Island Genovesa, however it is much shorter than *G. conirostris* and *G. magnirostris*. Hence, we expect that *G. difficilis* occupies a well differentiated niche space from the other two species. We tested these hypotheses on Genovesa and Española islands with morphometric data collected by Lack (1945, 1947): WingL, wing length; BeakH, Beak height; UbeakL, upper beak length and N-UBkL, nostril upper beak length. Data is available from Dryad Digital Repository: <https://doi.org/10.5061/dryad.150>. The original measurements (in mm) were log10-transformed prior to the construction of hypervolumes. Hypervolumes for each species were calculated using a gaussian kernel density estimator with the hypervolume R package with default parameters (see Blonder et al., 2018 for details). Then, hypervolume decomposition (Eqs 1–7) was carried with the function “kernel.beta” of the BAT package (Cardoso et al., 2015). Hypervolumes are reported in units of SDs to the power of the number of trait dimensions used.

To demonstrate how to estimate hypervolumes with a dataset with mixed type variables (continuous and non-continuous), a second analysis was performed. For the case, the WingL was transformed into a ranked ordered variable with 10 levels. Then, a Gower’s dissimilarity matrix was calculated and a NMDS was performed. The NMDS axes were used to estimate the hypervolumes for each species and to carry out hypervolume decomposition.

**RESULTS**

As hypothesized, hypervolumes calculated using morphometric data for *Geospiza conirostris* and *Geospiza magnirostris* were well differentiated on Genovesa Island (Figure 3), the niche overlap...
between these species being null ($H_{\text{overlap}} = 0$). By decomposing total differentiation ($H_{\text{total}} = 1$) into replacement and differences between niche breadths, we found that such differentiation was mostly caused by niche replacement ($H_{\text{repl}} = 0.609$) and not contraction/expansion, although the latter was also considerable ($H_{\text{diff}} = 0.391$). By comparing the niche of $G. \text{ conirostris}$ on Genovesa and Española, we found that they overlap slightly ($H_{\text{overlap}} = 0.246$) and the differentiation between both islands was mainly due to the replacement of niche space ($H_{\text{repl}} = 0.709$), not contraction/expansion ($H_{\text{diff}} = 0.045$). Therefore, the replacement component contributed 94% to the total differentiation, whilst differences in niche space contributed only 6% to total differentiation. This result reinforces the interpretation that the presence of $G. \text{ magnirostris}$ on Genovesa Island led to a significant shift (not contraction) of $G. \text{ conirostris}$ niche.

In relation to $G. \text{ difficilis}$, we found that its niche was highly differentiated from both $G. \text{ conirostris}$ ($H_{\text{total}} = 1$) and $G. \text{ magnirostris}$ ($H_{\text{total}} = 1$). The differentiation between $G. \text{ difficilis}$ and $G. \text{ conirostris}$ was caused, almost in equal proportions, by the replacement of niche space ($H_{\text{repl}} = 0.511$) and differences of niche breadth ($H_{\text{diff}} = 0.489$). However, the replacement of niche space was clearly the dominant component ($H_{\text{repl}} = 0.879$; $H_{\text{diff}} = 0.121$) of the differentiation between $G. \text{ difficilis}$ and $G. \text{ magnirostris}$.

The analyses carried out with mixed type variables (continuous and non-continuous) are presented in Supplementary Material. Results were similar to that obtained using only continuous data.

**DISCUSSION**

Hutchinson’s fundamental niche concept holds a central role across different fields of ecology and evolution (e.g., Holt, 2009; Blonder, 2018). Conceptually, the species niche corresponds to an $n$-dimensional hypervolume enclosing the range of conditions under which the species can survive and reproduce. Niche overlap occurs when species use the same resources or environmental conditions. In this regard, a crucial question is which processes determine the differentiation between species niches and, ultimately, determine species coexistence (MacArthur and Levins, 1967). In this paper, we propose a novel framework to partitioning overall differentiation between hypervolumes into two distinct fractions: niche shifts corresponding to the differentiation that results from the replacement of space between hypervolumes and niche contraction/expansion, corresponding to the differentiation that results from the net differences between the space enclosed by hypervolumes.

We illustrate our framework with a classic dataset (Lack, 1945, 1947). Our results revealed that differentiation between $G. \text{ conirostris}$ and $G. \text{ magnirostris}$ occurred mostly by niche shifts processes and not by the niche contraction of $G. \text{ conirostris}$.
on Island Genovesa, where both species live in sympatry. This means that *G. conirostris* occupied a different niche space and not a reduced space on Genovesa. These results are consistent with the hypothesis proposed by Lack (1947) that morphology of beaks was causally influenced by interspecific competition for food resources. This hypothesis received considerable support by analyzing the food habits of these species (Grant and Grant, 1982). These authors showed that *G. magnirostris* feeds almost entirely on large-hard seeds, whilst *G. conirostris* exploits Opuntia and arthropods on Genovesa. Moreover, on Española Island, where *G. magnirostris* is absent, *G. conirostris* consume large hard seeds. Therefore, it seems that the presence of *G. magnirostris* induces a shift in the diet of *G. conirostris* and, consequently, on its beak morphology.

As expected, our results showed that *Geospiza difficilis* was clearly differentiated from the other two species. However, distinct processes determined the differentiation of *G. difficilis*, from both *G. conirostris* and *G. magnirostris*. In relation to the former, the replacement of space (niche shifts) and differences of niche breadth (contraction or expansion) were both important, whilst in relation to the latter, the replacement component dominated. This result emphasizes the importance of considering the different components of differentiation to understand the processes that determine niche partitioning among species.

The functional roles and the sensitivity to environmental changes of the vast majority of taxa are usually unknown (the so-called Hutchinsonian shortfall; Cardoso et al., 2011). Therefore, a trait-based quantification of species niche can provide a practical way to assess niche parameters for a potentially large number of species, making it possible to understand and predict species niche responses to environmental changes (Violle and Jiang, 2009). Trait-based approaches have been used successfully to integrate functional ecology with community assembly and coexistence theories (e.g., McGill et al., 2006; Ackerly and Cornwell, 2007; Kraft et al., 2008). A major assumption of this approach is that trait dissimilarity is related to decreasing niche overlap among co-existent species (e.g., Stubs and Wilson, 2004), a rationale related to the limiting similarity principle (MacArthur and Levins, 1967). This premise assumes, implicitly, that the trait space is a good surrogate for the niche space. However, linking trait patterns to niche differentiation remains a challenge and, ultimately, depends on which traits were included in the analysis (D’Andrea and Ostling, 2016). Thus, an informed choice of traits should be made, based on their presumed adaptation to the species’ ecological niche.

Although, the framework proposed has been tested in the context of functional and community ecology, we advocate that it could be also a useful tool in other areas of ecological niche modeling (e.g., Blonder, 2018; Mammola, 2019). For example, bioclimatic hypervolumes can be constructed and compared between species or within the same species across different geographic areas or time periods. This may allow to identify which processes determine niche changes (shifts or contractions/expansion of niche space) along climatic gradients or in response to climatic changes between time periods.

In conclusion, an integrated framework was proposed to partition overall differentiation between hypervolumes into distinct fractions, replacement of space (niche shifts) and net differences between space amplitudes (niche contraction/expansion processes). This framework allows quantifying the relative importance of each process in determining the differentiation of species niches. In that sense, the method offers novel insights to understand the drivers of niche partitioning strategies among coexistence species. We expect that the methods here presented will allow to address a wider range of niche- and trait-based questions than was possible to date.

**DATA AVAILABILITY STATEMENT**

The datasets generated for this study are available on request to the corresponding author.

**AUTHOR CONTRIBUTIONS**

All authors listed have made a substantial, direct and intellectual contribution to the work, and approved it for publication.

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**SUPPLEMENTARY MATERIAL**

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fevo.2020.00243/full#supplementary-material

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