

1 **Diverse strategies for tracking seasonal environmental niches at hemispheric scale**

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14 **Author statement:** All authors agreed to submission of the manuscript and accept the
15 responsibility for the accuracy and integrity of the manuscript.

16

17 **Author contributions:** All authors devised ideas for the study. JMC compiled annotated GBIF
18 data. JMC calculated niche similarity metrics. JMC accessed functional trait databases and
19 harmonized phylogenetic information. JMC conducted statistical analyses and generated figures.
20 All authors helped write the manuscript.

21

22 **Keywords:** Environmental niche, climatic niche, hypervolume, seasonal niche tracking, birds,
23 big data, seasonality, migration, functional trait groups, phylogenetic signal, GBIF

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25

26 **Abstract:**

27

28 Species depend upon a constrained set of environmental conditions, or niches, for survival and
29 reproduction that are increasingly lost under climatic change. Seasonal environments require
30 species to either track their niches via movement or undergo physiological or behavioral changes
31 to survive. Here we identify the tracking of both environmental niche position and breadth across
32 619 New World bird species and assess their phylogenetic and functional underpinning.
33 Partitioning niche position and breadth tracking can inform whether climatic means or extremes
34 limit seasonal distributions. We uncover diverse strategies, including the tracking of niche
35 position, breadth, both, or neither, suggesting highly variable sensitivity to ongoing climatic
36 change. There was limited phylogenetic determinism to this variation, but a strong association
37 with functional attributes that differed between niche position and breadth tracking. Our findings
38 imply significant functional consequences for communities and ecosystems as impending climate
39 change affects some niche tracking strategies more than others.

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41 Introduction

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43 Species survive and reproduce under a specific set of environmental conditions, known as the
44 environmental niche or n -dimensional hypervolume (Blonder et al. 2014; Hutchinson 1957; Lu,
45 Winner, and Jetz 2021). In seasonal environments, species must adjust to a constantly shifting
46 window of available conditions through one of several strategies. When remaining stationary,
47 species must either maintain tolerance to a wide range of conditions or undergo physiological
48 and behavioral changes to survive seasonal variation, known as ‘niche switching’ (Nakazawa et
49 al. 2004). Alternatively, seasonally mobile animals can occupy a dynamic niche that remains
50 relatively narrow across the annual cycle, known as ‘niche tracking’ (Gómez et al. 2016;
51 Somveille, Rodrigues, and Manica 2018; Winger et al. 2019). Seasonal niche tracking is central
52 to the persistence of species with limited behavioral or physiological capacity to adjust their
53 niches (Fandos et al. 2020; Zurell et al. 2018). Understanding the functional and phylogenetic
54 drivers of niche tracking behavior across diverse species can allow researchers to predict how
55 species mediate exposure to novel, potentially adverse conditions as climate change progresses
56 (Tingley et al. 2009; La Sorte and Jetz 2012; Somveille, Rodrigues, and Manica 2015).

57 After the documentation of seasonal niche tracking behavior in single species (Fandos et al.
58 2020) and smaller clades (Gómez et al. 2016; Eyres et al. 2020), a more general understanding of
59 the patterns, causes and consequences of seasonal niche tracking across diverse taxonomic
60 groups remains missing. For example, the role of phylogeny and functional traits in driving
61 seasonal niche tracking across a diverse species set remains largely unexplored (but see Zurell et
62 al. 2018). Several studies have hypothesized that niche tracking may be phylogenetically
63 conserved (Gómez et al. 2016; Martínez–Meyer, Townsend Peterson, and Navarro–Sigüenza
64 2004), as is typical of behavioral and migratory traits (Outlaw and Voelker 2006), but this has
65 not been evaluated rigorously or broadly. Alternatively, niche tracking may have repeatedly
66 evolved in tandem with species’ functional traits. For example, obligate insectivores or small-
67 bodied species may be most likely to closely track their niche over the annual cycle to satisfy
68 narrow dietary or thermal requirements (Gómez et al. 2016; Huey et al. 2012). Separating these
69 potential drivers will allow researchers to better predict niche flexibility and niche tracking, and
70 thus climate change vulnerability, among rare species or those from under-sampled regions.

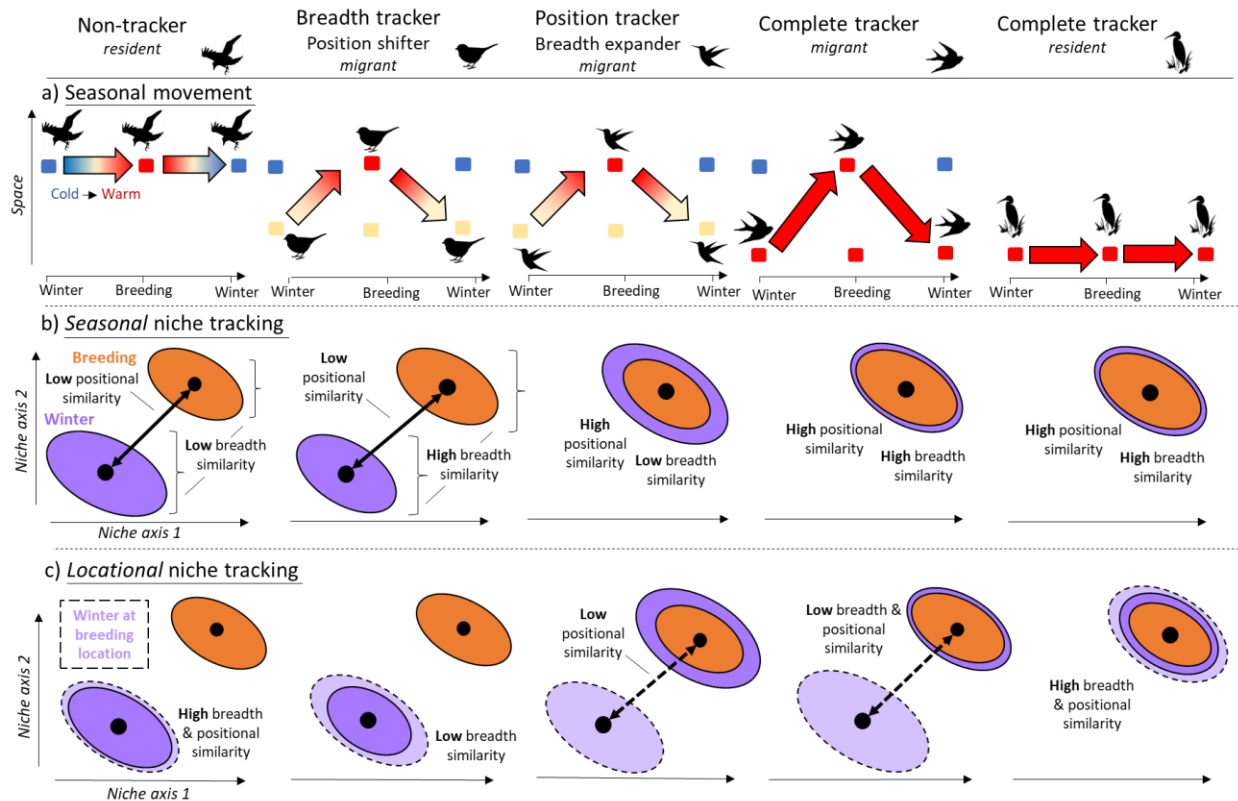
71 Previous work on niche tracking has focused on seasonal similarity in niche positions, or mean
72 environmental conditions, without accounting for niche components such as niche breadth (the
73 volume or range of tolerable conditions). However, niche breadth is a central additional
74 dimension because climate change is altering both the means and variances of climatic
75 conditions (Rahmstorf and Coumou 2011). For example, species may seasonally track niche
76 breadth instead of changing the central niche position when extremes are more limiting to their
77 survival and reproductive success than climatic means, as is the case for numerous species
78 (Albright et al. 2010; Ummenhofer and Meehl 2017). Thus, species may use the tracking of
79 niche breadth as an alternative strategy to niche position tracking to persist in the face of

80 seasonality, one unexplored by the existing niche tracking literature, which instead considers
81 niche breadth as an annually static variable (Gómez et al. 2016; Zurell et al. 2018). To better
82 identify complex variation in niche tracking, seasonal niche similarity should be partitioned into
83 constituent components, including *niche position similarity*, or the distance between niche
84 centroids, representing the difference between the average conditions a species experiences in
85 each season; and *niche breadth similarity*, or the proportional difference in niche breadths (Lu,
86 Winner, and Jetz 2021), representing the range of conditions a species can tolerate.
87 Hypervolumes quantified using non-parametric techniques, such as kernel-density estimates or
88 support vector machines (Blonder 2018; Brown, Holland, and Jordan 2020), are difficult to
89 partition into constituent components. However, recently developed parametric methods for
90 quantifying the niche allow for partitioning of these components and direct hypothesis testing
91 against predictions derived from theories (Lu, Winner, and Jetz 2021).

92 We use these new metrics to distinguish five primary strategies for seasonal niche tracking (Fig.
93 1; Fig. 2): i) “Non-trackers” retain neither the position nor the breadth of their niches over the
94 annual cycle; ii) “Breadth trackers”, or “position shifters”, track niche breadth but shift their
95 position, suggesting that their seasonal ranges may be limited by environmental variation; iii)
96 “Position trackers”, or “breadth expanders”, track only niche position and adjust niche breadth
97 between seasons and may be seasonally limited by environmental means; Finally, “complete
98 trackers” track both niche position and breadth, but may do so by either iv) migrating to track
99 weather conditions across the annual cycle or v) remaining stationary in an aseasonal
100 environment.

101 However, a simple comparison of niche space during the breeding and overwintering seasons
102 (henceforth, *seasonal* niche similarity) cannot fully distinguish these strategies (Fig. 1b, columns
103 4-5), despite obviously divergent dispersal capability and potential for climate change adaptation
104 (Eyres et al. 2020). We therefore consider an additional metric, *locational* niche similarity, in
105 which the realized overwintering hypervolume is compared with that had the species remained at
106 its breeding range, its *stationary winter niche* (Fig. 1c). This measure, which emphasizes the
107 similarity of conditions at specific locations, is important for understanding seasonal niche
108 tracking because it accounts for the environmental distance covered by the species through
109 purposeful movement. Among temperate breeders, seasonal and locational niche similarity are
110 likely to have an inverse relationship; for example, a niche tracking species with little difference
111 between its breeding and overwintering niche is likely to experience highly distinct
112 overwintering conditions compared with those at its breeding sites.

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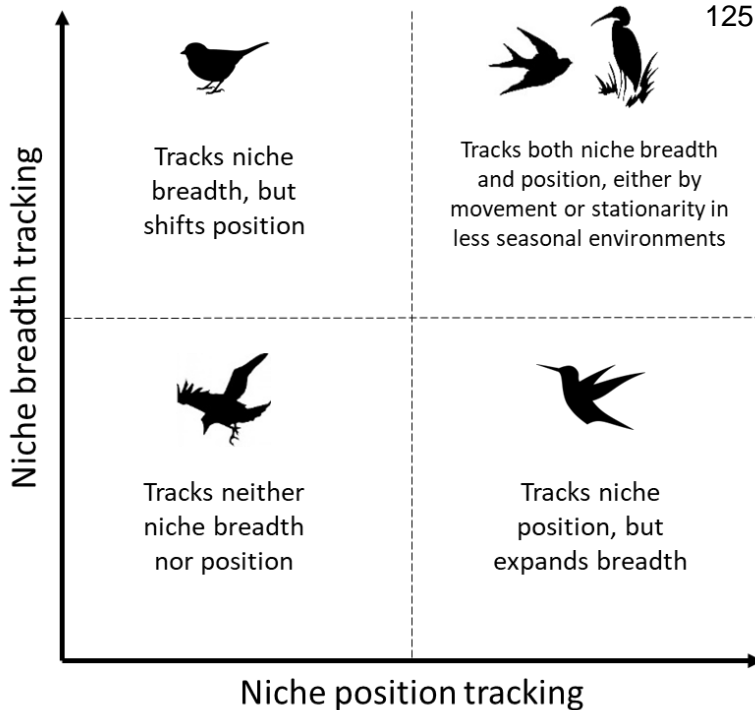
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Figure 1: A typology of seasonal niche tracking strategies. Seasonal niche similarity is decomposed into two components, similarity in niche position and niche breadth. A given species may (1) be a non-tracker, tracking neither niche position or breadth, (2) track niche breadth, but shift position, (3) track niche position, but expand breadth, or (4-5), be a complete tracker, tracking both niche position and breadth. a) shows the differing use of climate zones over space and time; b) compares seasonal niches between the breeding (orange) and overwintering (purple) seasons (solid arrows represent positional similarity, and brackets represent breadth similarity); c) compares the location of the winter conditions at breeding locations (light purple) and the overwintering niche, revealing that for strategy 4 and 5 niche tracking is achievable through multiple life history strategies. Dashed arrows represent locational positional similarity.



126

127 **Figure 2. Five niche tracking strategies.** Conceptual schematic outlining the contribution of
128 niche breadth and position tracking to each of five possible niche tracking strategies.

129

130 Here, we use a hemispheric system to address patterns and drivers of these strategies;
131 specifically, 619 bird species, representing nearly the full diversity of birds breeding in the US
132 and Canada. This species set is ideally suited given its tremendous variation in migratory
133 strategy, diversity of functional trait groups (Barnagaud et al. 2017) and uniquely comprehensive
134 occurrence data. We also leverage a recently developed environmental annotation tool (Li et al.
135 2021) and a novel parametric hypervolume method (Lu, Winner, and Jetz 2021) to quantify
136 similarity between the environmental niches occupied by each species during both the breeding
137 and overwintering seasons. We use this system to address the following questions:

138 1) What is the prevalence of different niche tracking strategies while accounting for both niche
139 position and breadth? To date, a quantitative multi-species assessment of strategies across a
140 diverse taxonomic group has been missing.

141 2) What is the relationship between the tracking of niche position and the tracking of niche
142 breadth? Do species that maximize niche positional similarity across seasons also tend to retain
143 similar niche breadth, or do species largely track only one or the other?

144 3) How phylogenetically and functionally determined are niche tracking strategies? Across
145 functional trait groups, we hypothesize seasonal niche tracking to be most common in a) long-
146 distance migrants, because they can physically relocate to suitable locations (Laube, Graham,
147 and Böhning-Gaese 2015; Zurell et al. 2018); b) small-bodied species, because they have low
148 thermal inertia and generally narrow thermal breadths (Huey et al. 2012; Albright et al. 2017),
149 suggesting they cannot tolerate large seasonal variation in the niche; c) insectivores, because
150 abundant insect prey is only available under specific temperature, precipitation and productivity
151 levels (Winkler, Luo, and Rakhimberdiev 2013); and d) species occupying open or water
152 habitats, because they are not shielded from climate variability by forest structure (Jarzyna et al.
153 2016). Given that behavioral and migratory traits are often phylogenetically conserved (Outlaw
154 and Voelker 2006), we expect a strong phylogenetic signal in seasonal niche tracking behavior
155 (Gómez et al. 2016).

156 4) Do cross-species comparisons of *locational* niche similarity reveal important behavioral
157 strategies and functional or phylogenetic associations not apparent when quantifying only
158 *seasonal* niche similarity? We predict an inverse relationship between seasonal and locational
159 similarity across species and expect to observe functional trait relationships with locational
160 similarity that are not observed with temporal similarity.

161 We expect the emerging insights to not only address these questions but more generally offer an
162 assessment of seasonal niche dynamics as a system to understand realized strategies for
163 mitigating exposure to climatic change.

164

165 **Material and Methods**

166

167 *Species selection and environmental data*

168 We identified the 672 bird species that annually breed or overwinter in the United States and
169 Canada based on American Birding Association birding codes (Association 2008) updated to the
170 Clements bird taxonomy as of 2021 (Clements 2007). These codes are a widely-accepted
171 authority to distinguish regularly occurring species from irregularly occurring vagrants, or those
172 species which sporadically appear on the continent each year but whose occurrence is not
173 predictably tied to a given location. We excluded species not native to the US or Canada or those
174 that are primarily marine, for which weather data is unavailable.

175 In August 2021, we accessed the Spatiotemporal Observation Annotation Tool (STOAT) v1.0, a
176 novel cloud-based toolbox for flexible biodiversity annotations (Li et al. 2021), to download
177 annotated Global Biodiversity Information Facility data (<https://www.gbif.org/>) for all species
178 (data compilation, analyses and visualizations were all completed in R 4.1.0; R Core Team
179 2021). To minimize the potential for spatiotemporal sampling bias, we then thinned points by
180 selecting one point from each location (5x5km grid cell) per week.

181 We annotated observation points with three environmental dimensions: daily maximum
182 temperature (sourced from NASA-MODIS; <https://lpdaac.usgs.gov/products/mod11a1v006/>),
183 enhanced vegetation index (EVI; from MODIS), and precipitation (from CHELSA v2.1; Karger
184 et al. 2021), each summarized to a 1km buffer (0.5km radius) over 30 days prior to the
185 observation (imported using jsonlite and httr packages; Ooms 2014; Wickham and Wickham
186 2020). We did not assess *a priori* whether these variables are equally relevant across species; for
187 example, certain species may be limited by precipitation but not temperature within their range,
188 or *vice versa*. Alternatively, some species may be limited by environmental factors not
189 considered in our estimation of species niches. However, this variable set represents the
190 environmental factors that most commonly drive species distributions (Qian 2010), and these
191 variables have low collinearity, allowing each axis to remain independent. Estimating niches
192 with the same variable set for all species was necessary to ensure consistency in cross-species
193 evaluation of seasonal changes to both niche position and breadth.

194 For each species, we restricted observation points to those with environmental data available for
195 all dimensions. Further, we temporally cropped data to season (December-February for the
196 overwintering season and June-August for breeding season) and spatially cropped points to the
197 American continents ($< -30^\circ$ longitude). Sufficient data for analysis (>20 points per season after
198 filtering) was available for 619 species. Database management was completed using *tidyverse*
199 packages (Wickham et al. 2019).

200 To quantify locational similarity, we conducted the same niche characterizations for breeding
201 locations during the overwintering season. For each species, we created a hypothetical set of
202 occurrence points corresponding to coordinate locations during the breeding season, each with
203 randomized winter calendar dates (December-February), annotating them with environmental
204 data as described above.

205 *Ecological niche modeling and seasonal similarity*

206 For every species, we calculated parametric measurements of the similarity between the seasonal
207 three-dimensional hypervolumes. The hypervolumes were characterized as multivariate normal
208 distributions, allowing us to derive analytical estimates for the breadth and position of each
209 hypervolume (MVNH package; Lu, Winner, and Jetz 2021). We partitioned niche similarity into
210 two metrics: i) *niche position similarity*, which quantifies the distance among hypervolume
211 centroids in each season based on the sign-flipped, log-transformed Mahalanobis distance, used
212 to assess changes in niche position between seasons; and ii) *niche breadth similarity*, which
213 represents the similarity in *niche breadth* (volume) between seasons and is measured as the sign-
214 flipped, log-transformed determinant ratio.

215 To quantify seasonal similarity, we calculated positional and breadth similarity between the
216 breeding and overwintering seasonal hypervolumes. To quantify locational similarity, we
217 calculated positional similarity between the overwintering hypervolume and available

218 hypervolume during winter at breeding locations. Thus, we had a total of three metrics of
219 seasonal niche similarity per species. We visualized pairwise two-dimensional hypervolumes for
220 each species using *ggplot2* (Wickham 2011).

221 *Functional traits and phylogeny*

222 We obtained species-level functional trait values and an avian phylogenetic tree to test
223 associations between trait groups, phylogeny, and seasonal niche tracking. We derived migration
224 distances from (La Sorte et al. 2022; La Sorte personal communication), body mass from the
225 Eltonian trait database (Wilman et al. 2014), and habitat preference and diet from Barnagaud et
226 al. (2017). We grouped several categories of each categorical predictor to avoid false positives
227 associated with small sample sizes and to keep our conclusions broad. Diet categories included
228 carnivore, invertebrate, omnivore, and herbivore (combining ‘fruit’, ‘nectar’, ‘vegetation’, and
229 ‘seed’ categories). Habitat categories included water (‘coastal’, ‘open_water’,
230 ‘riparian_wetlands’), open (‘semi-open’, ‘rock’, ‘arid’), generalist (‘urban’, ‘developed’), and
231 forest.

232 We updated an avian phylogeny from Jetz et al. (2012) to account for recent taxonomic changes,
233 updating species names to the Clements bird taxonomy as of 2021 (Clements 2007) and treating
234 recently split species as having no phylogenetic distance (Appendix 2). We also harmonized
235 species names in all trait datasets to Clements.

236 *Cross-species models*

237 To assess the role of functional traits in seasonal niche tracking, we used weighted multivariate
238 phylogenetic generalized least squares (PGLS) models (*caper* package; Orme et al. 2013).
239 Phylogenetically correlated model errors in PGLS account for the non-independence of the
240 species due to their phylogenetic relatedness (Symonds and Blomberg 2014). The dependence of
241 the model errors arises from trait axes that we did not include in the analysis and that may be
242 subject to niche conservatism so that model errors reflect the unobserved trait and thus the
243 phylogenetic distance between species. We used species’ similarity in niche position and breadth
244 as the response variables in the models and explained their variation with categorical trait
245 variables. Prior to use in models, continuous predictor variables (migration distance and log-
246 transformed body mass) were scaled to improve model fit and response variables were log-
247 transformed and sign-flipped (representing niche similarity rather than dissimilarity) to improve
248 interpretability. We weighted all species points in models by log-transformed minimum sample
249 size (number of points in the season with less data) to account for uncertainty in seasonal niche
250 dissimilarity estimates (true error estimates are unavailable when estimating niche dissimilarity).
251 We fit three models testing the relationships between the four functional traits and three metrics
252 of niche similarity: seasonal niche position similarity, seasonal niche breadth similarity, and
253 locational niche position similarity. To evaluate the reliability of multivariate models, we also fit
254 univariate models for each combination of functional trait predictor and response variable.

255 To measure the role of phylogeny in both positional and breadth similarity, we quantified
256 Blomberg's K (Blomberg, Garland Jr et al. 2003) and compared it to a null distribution of K after
257 randomizing species' responses 1,000 times (picante package; Kembel et al. 2010). $K < 1$
258 suggests greater than expected phylogenetically-correlated variance within clades, while $K > 1$
259 suggests variance among clades (i.e., phylogenetically-conserved trait). We also quantified
260 lambda to assess the extent of phylogenetic conservation of niche metrics via Brownian motion.
261 We visualized relationships between niche metrics and functional traits using *ggplot2* (Wickham
262 2011), *ggExtra* (Attali and Baker 2019) and *RcolorBrewer* (Neuwirth and Neuwirth 2011). We
263 visualized partial residuals using *visreg* (Breheny, Burchett, and Breheny 2020).

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265 **Results**

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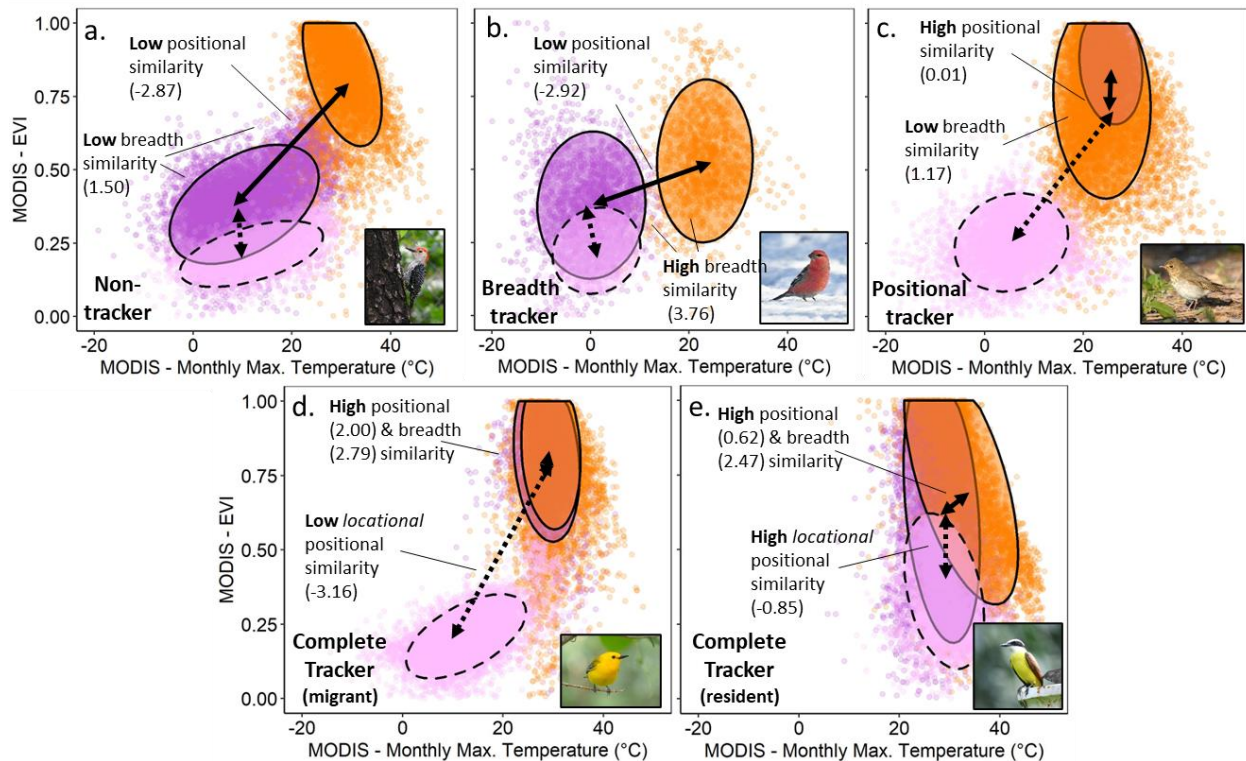
267 We found all five niche-tracking strategies represented among North American birds (Fig 3). For
268 example, Red-bellied woodpeckers (*Melanerpes carolinus*; Fig. 3a) are largely stationary birds
269 occupying highly seasonal environments, with low seasonal similarity between their breeding
270 and overwintering niches, but high locational similarity between their overwintering and
271 stationary winter niches. In contrast, Pine grosbeaks (*Pinicola enucleator*; Fig. 3b) are short-
272 distance migrants and “breadth trackers”, tracking niche breadth while shifting position across
273 seasons. Migratory Swainson's thrushes (*Catharus ustulatus*; Fig. 3c) are “position trackers”,
274 tracking their niche position across seasons, but expanding their niche during the breeding
275 season. Prothonotary warblers (*Protonotaria citrea*; Fig. 3d) are long-distance migrants that
276 abandon breeding territories to closely track both niche components across seasons, resulting in
277 high seasonal and low locational similarity. Finally, Great kiskadees (*Pitangus sulphuratus*; Fig.
278 3e) are stationary tropical residents with conserved niches across the annual cycle, resulting in
279 both high seasonal and high locational similarity. We select these species for visualizations
280 because they represent each strategy well in both two-dimensional niche space (temperature and
281 EVI only; Fig. 3) and three-dimensional space (remainder of results).

282

283 The 619 bird species analyzed spanned the full diversity of niche tracking strategies (Fig. 4;
284 Appendix 1). Across seasons, the median positional similarity ($-\log[\text{Mahalanobis distance}]$) was
285 -1.95 ± 0.05 SE (i.e., centroids approximately 3 SDs apart), suggesting that many species are
286 not closely tracking this niche attribute over the annual cycle (Ponti et al. 2020). For 60 species
287 (9.7%), centroids were within 1 SD, suggesting highly similar niches in each season for this
288 subset. The median among-season similarity in niche breadth was 2.17 ± 0.03 , equivalent to a
289 4:1 ratio in niche breadth between seasons, suggesting that many species expand their niche
290 breadths as they traverse seasons. 129 species (22%) had a niche breadth similarity equivalent to
291 a $< 2:1$ ratio, suggesting similarly-sized seasonal niches.

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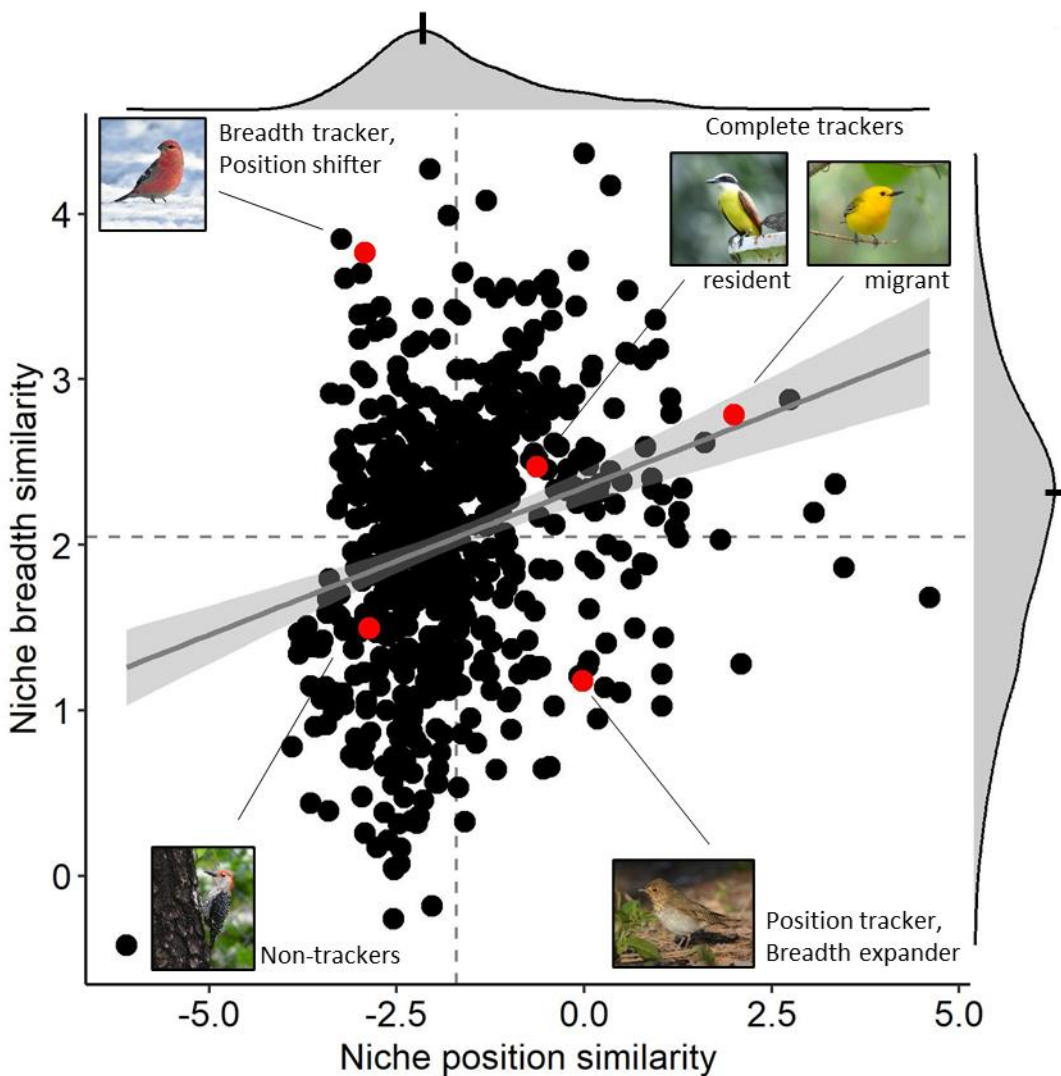
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295 **Figure 3: Diverse niche tracking strategies.** (a) Red-bellied woodpeckers (*Melanerpes carolinus*) are
 296 non-trackers, with low seasonal similarity (solid arrow) between their breeding (orange points and
 297 ellipsoid) and overwintering (dark purple) two-dimensional niche (presented as standardized ellipse area).
 298 Meanwhile, the difference between their overwintering niche and winter conditions at breeding locations
 299 (light purple, dotted ellipsoid), or locational similarity (dotted arrow), is high. (b) Pine grosbeaks
 300 (*Pinicola enucleator*) track their niche breadth but shift position across seasons, while (c) Swainson's
 301 thrushes (*Catharus ustulatus*) track their niche position but expand niche breadth. (d) Prothonotary
 302 warblers (*Protonotaria citrea*) closely track their niche across seasons via migration, resulting in high
 303 seasonal and low locational similarity. (e) Great kiskadees (*Pitangus sulphuratus*) track their niche by
 304 remaining stationary in less seasonal environments, leading to high seasonal and locational similarity.
 305 Position and breadth similarity values are provided in parentheses. Note that this figure displays two-
 306 dimensional niche space, though our statistical analyses consider three-dimensional niches, including
 307 precipitation as additional dimension.

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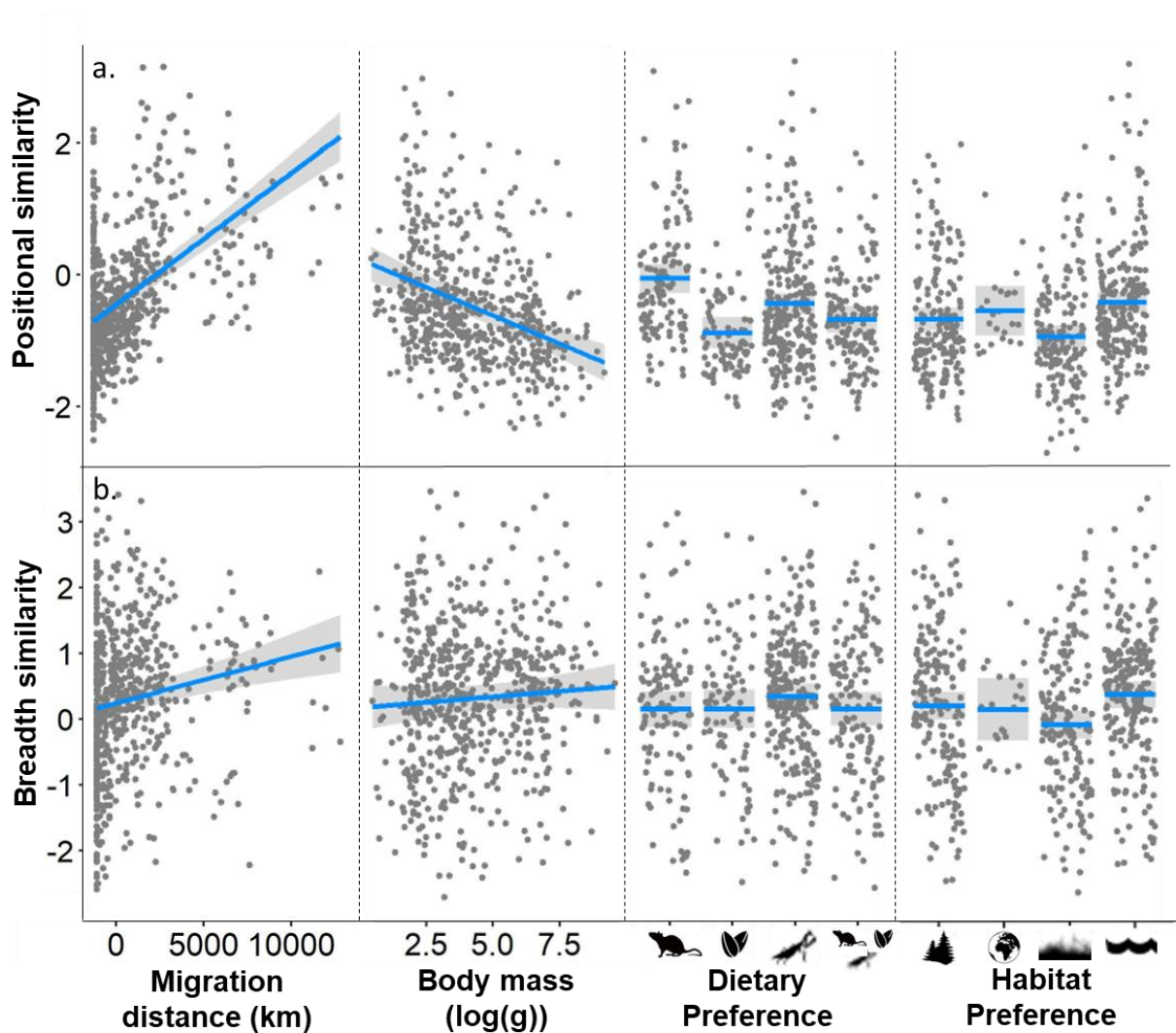
309 Tracking niche position was loosely tied to tracking niche breadth; species that tracked niche
 310 position more often tracked niche breadth, and *vice versa* (Generalized linear model: $\beta=0.179$,
 311 $R^2=0.073$, $p<0.001$; Fig. 4). 170 (27.5%) species were "complete" niche trackers, tracking both
 312 position and breadth more than most species (positive values on both axes in Fig. 4), while 209
 313 (33.8%) were non-trackers, tracking both less than others (negative values in Fig. 4). However,
 314 numerous species tracked one but not the other (Fig. 4), suggesting a complex diversity of
 315 approaches to seasonal niche tracking across bird species. Only 80 species (12.9%) tracked niche

316 position but expanded breadth, while double that number – 160 (25.8%) – tracked breadth but
317 shifted position.
318



319
320 **Figure 4: Interspecific variation in tracking niche position and breadth across seasons.** The figure
321 shows niche similarities between the breeding and overwintering seasons for the 619 bird species
322 analyzed based on three niche dimensions (for red highlights, see Fig. 3). Strategies of tracking niche
323 position, breadth, both, or neither are partitioned into four categories (boxes separated by dashed lines).
324 Dashed lines highlight the means. Marginal density plots illustrate that species tracking only one niche
325 component often track niche breadth more closely than niche position, with medians represented by
326 notches. The trendline represents a linear relationship between niche position and breadth similarity with
327 associated 95% confidence interval (shading). See Fig. S2 for equivalent patterns for locational similarity.
328
329 We found seasonal niche tracking strategies to be closely associated with functional traits, and
330 their role to be consistent across weighted multivariate models accounting for all traits and
331 phylogenetic structure simultaneously (Fig. 5; Tables S1-S2) as well as univariate models

332 (Tables S3-S4). As expected, both niche positional and breadth similarity were very strongly
333 linked to migration distance (Phylogenetic generalized least-squares models: position, $\beta=0.43$,
334 $p<0.001$; breadth, $\beta=0.25$, $p<0.001$), which was the strongest functional predictor; long-distance
335 migrants were most likely to maximize similarity, while residents minimized similarity and
336 others fell in the middle (Fig. 5a,d). Body mass emerged as another important functional
337 predictor of seasonal niche tracking, especially with regards to positional similarity. Body mass
338 negatively predicted niche position similarity ($\beta=-0.80$, $p<0.001$) and was positively associated
339 with breadth similarity ($\beta=0.53$, $p<0.001$). Small-bodied birds maximized seasonal similarity,
340 while large-bodied birds minimized it (Fig. 5b,e). Finally, both diet ($F_{1,3}=5.70$, $p<0.005$) and
341 habitat preference ($F_{1,3}=9.68$, $p<0.001$) categories described variation in positional similarity,
342 though only habitat preference drove breadth similarity ($F_{1,3}=10.15$, $p<0.005$). As predicted,
343 insectivores maximized positional and breadth similarity, herbivores minimized these, and
344 omnivores fell in the middle (Fig. 5c,g). Waterbirds tracked their niches more closely than other
345 species (Fig. 5d,h).
346



347

348 **Figure 5: Functional traits predict variation in seasonal niche tracking**, as shown by partial residual
349 plots based on phylogenetic least-squares models. Y-axis values represent (a) positional and (b) breadth
350 similarity. Silhouettes correspond to categorical functional trait values as follows: Dietary preference (left
351 to right: carnivore, herbivore, insectivore, omnivore); Habitat preference (forest, generalist,
352 open/grassland, water).

353

354 Surprisingly, we found little evidence of a phylogenetic signal in seasonal niche tracking. For
355 both niche position and breadth, Within-group phylogenetic variation was greater and among-
356 group variation lesser than expected by chance (Bloomberg's K always < 0.1; position: $p < 0.001$;
357 breadth: $p = 0.10$; Table S5) and did not support Ornstein-Uhlenbeck trait evolution.

358 Measurements of lambda indicated a moderate phylogenetic signal for species tracking niche
359 position ($\lambda = 0.66$, $p < 0.001$), suggesting possible Brownian-motion trait evolution, but little signal
360 for niche breadth ($\lambda = 0.28$, $p < 0.001$).

361

362 Species that maximized seasonal niche similarity typically minimized locational niche similarity,
363 and vice versa ($\beta=-0.408$, $R^2=0.237$, $p<0.001$; Fig. S1). For locational similarity, the link
364 between tracking niche position and niche breadth was stronger than for seasonal similarity, with
365 42% of species tracking both components and 29% neither (Fig. S2). Only 15% of species
366 tracked locational niche position but not breadth, and 13% tracked locational breadth but not
367 position. Likewise, relationships between functional traits and locational positional similarity
368 were generally opposite to those observed with seasonal positional similarity. Long-distance
369 migrants ($\beta=-0.331$, $p<0.001$), small-bodied birds ($\beta=0.95$, $p<0.001$), and waterbirds ($F_{1,3}=7.75$,
370 $p<0.001$) each minimized locational positional similarity, though we found no effect of diet. As
371 with seasonal similarity, we found no evidence for a phylogenetic signal in locational niche
372 similarity (Table S5).

373

374

375 **Discussion**

376 For mobile organisms, seasonal niche tracking is a central part of the behavioral repertoire for
377 mediating exposure to novel climate conditions. Here, we assessed the representation of distinct
378 seasonal niche tracking strategies across North American breeding birds and evaluated the
379 functional and phylogenetic drivers of these strategies. Our analyses revealed five distinct
380 seasonal niche tracking strategies and their key functional drivers. We find that only a bit more
381 than half of species can be categorized as either “complete trackers”, tracking both niche position
382 and breadth across seasons, or “non-trackers”, tracking neither. However, almost 40% of species
383 only track niche position or breadth, revealing complexity in niche tracking strategies that is
384 previously unexplored in niche-tracking studies. Twice as many species tracked only niche
385 breadth (26%) compared with those that tracked niche position and expanded niche breadth
386 (13%), suggesting that tracking breadth is often a prerequisite for tracking position and that birds
387 may be most specialized in terms of the range of environmental conditions they can tolerate
388 rather than means. Importantly, niche tracking studies that do not consider niche breadth would
389 likely have labeled the 26% of species that are niche breadth trackers as non-trackers, ignoring
390 the role that niche breadth tracking and expansion plays in allowing species to persist over
391 seasons. Species limited by extreme or highly variable weather conditions rather than climatic
392 means may be more apt to prioritize niche breadth tracking while allowing niche position to
393 seasonally fluctuate. Meanwhile, a species with inflexible thermal or dietary requirements that is
394 specialized only during certain parts of the year (e.g., due to strict breeding requirements) may
395 track niche position but not breadth. However, further work is needed to pinpoint the ecological
396 motivations behind selection of breadth tracking as a niche tracking strategy.

397 We found that migration distance was the most important functional driver of seasonal niche
398 tracking, despite this link previously receiving mixed support among cross-species niche tracking
399 studies (Zurell et al. 2018; Laube, Graham, and Böhning-Gaese 2015; Gómez et al. 2016).

400 Although several studies (Gómez et al. 2016; Laube, Graham, and Böhning-Gaese 2015) failed

401 to detect a link between migration and niche tracking, their approaches compared species within
402 single families, including residential species that are exclusively tropical. Because these residents
403 occupy relatively stable environments, they can track their niches despite being stationary (Eyres
404 et al. 2020). This example helps illustrate why migration distance and niche tracking are less
405 correlated than might be expected – tropical residents can track their niches well despite their
406 stationarity, while long-distance migrants may experience high seasonal variation despite their
407 migration if they annually move between arctic breeding grounds and boreal or temperate
408 wintering grounds. Indeed, although migration distance was the strongest correlate of seasonal
409 niche tracking in our study, the correlation was moderate ($R^2=0.249$).

410
411 Body size was revealed as the next most important functional trait associated with seasonal niche
412 tracking when controlling for other traits. Small-bodied birds may not be able to tolerate large
413 variation in their thermal niches because they have low thermal inertia (Huey et al. 2012;
414 Albright et al. 2017). Further, small-bodied animals have shown greater sensitivity to extreme
415 weather events (Cohen, Fink, and Zuckerberg 2021) and more rapidly alter phenological timing
416 in tune with interannual weather variability (Cohen, Lajeunesse, and Rohr 2018) compared with
417 large-bodied animals, suggesting that they must carefully maintain thermal limits. Additionally,
418 small-bodied birds may be more likely to evolve niche-tracking strategies dependent on seasonal
419 migration because long-distance flight carries lower energetic costs than it does for large birds
420 (Watanabe 2016).

421
422 Finally, diet and habitat preference were additional meaningful drivers of seasonal niche tracking
423 behavior, with insectivores and carnivores tracking their niches more than herbivores and
424 omnivores, and waterbirds more so than other species. Insectivores may prioritize niche-tracking
425 because they are dependent upon prey that is most commonly available under specific thermal
426 and productivity limits (Winkler, Luo, and Rakhimberdiev 2013). Surprisingly, carnivores
427 tracked their niche positions even more closely than insectivores, but many of these species are
428 preying on herpetofauna also active during warm weather. Though we predicted both open and
429 water habitat specialists to be the strongest niche trackers, as they are less shielded from the
430 external environment by habitat structure (Jarzyna et al. 2016), only water birds tracked their
431 niche more closely than other groups (Fig 4d,h). Bird species that occupy open habitats may
432 make greater use of microhabitat structure than water birds, which are typically exposed (e.g.,
433 (Shew, Nielsen, and Sparling 2019)).

434
435 Although many studies have assumed seasonal niche tracking to be phylogenetically conserved
436 (Gómez et al. 2016; Martínez–Meyer, Townsend Peterson, and Navarro–Sigüenza 2004), this
437 has not to our knowledge been evaluated over a large, diverse taxonomic group such as North
438 American breeding birds. There are several explanations as to why we did not detect a
439 phylogenetic signal. First, migratory behavior has arisen in numerous, diverse lineages of birds,
440 and is flexible even within species (Zink 2011). Second, closely related species often occupy

441 distinct climatic zones that vary greatly in seasonality, especially if speciation resulted from
442 character displacement (Newton 2003). Finally, species-level exposure to and adaptation
443 potential for climate change may be quite independent from phylogeny (Khaliq et al. 2015;
444 Davis et al. 2010). Based on our analyses, we conclude that functional traits likely play a larger
445 role than phylogeny in driving seasonal niche tracking in birds.

446
447 Complex variation among niche tracking strategies cannot fully be assessed via seasonal niche
448 comparisons; for example, a species that niche-tracks via movement and a stationary species in
449 an aseasonal environment have similar seasonal niches. To parse these strategies, we estimated
450 locational niche similarity, or the similarity between overwintering niches and the *stationary*
451 *winter niche* for each species. Species with high seasonal similarity often had low locational
452 similarity, and *vice versa*, reflecting a trade-off between conserving niches and ranges
453 throughout the annual cycle. While many species tracked only niche position or breadth between
454 seasons, species adhered to locationally tracking both or neither components much more often,
455 with few tracking just one. Therefore, species may be leaving breeding areas in winter for the
456 purposes of modifying both the mean and variance of environmental conditions. We observed
457 several key differences between functional trait relationships and either seasonal or locational
458 niche similarity. For instance, migration distance and body size were equally important as drivers
459 of locational niche similarity, possibly because large-bodied birds are especially reluctant to
460 energetically invest in movement (Watanabe 2016). Further, in contrast with seasonal niche
461 similarity, there was no effect of diet on locational similarity, perhaps because food availability
462 depends more on environmental conditions themselves rather than the distance birds travel to
463 track them. Thus, we demonstrate the importance of considering both seasonal and locational
464 similarity in species niches to understand variation in niche tracking strategies.

465
466 We estimated niche metrics at the species level to effectively compare niche tracking strategies
467 across a taxonomic group as broad and diverse as birds and to make use of the enormous
468 quantity of available species-level occurrence data. However, we did not explore niche variation
469 as experienced at lower levels of organization, including populations, demographic groups, and
470 individuals (Fandos et al. 2020; Fandos and Tellería 2020; Carlson et al. 2021). Populations of
471 species with wide geographic ranges might be adapted to highly distinct environmental
472 conditions (Broggi et al. 2005). Within a population, demographic groups can also occupy
473 different niches – sexes can occupy spatially distinct regions during the nonbreeding season, or
474 young can be shielded in microclimate during breeding season (Shipley et al. 2020). Finally,
475 individuals may track different niche components than populations (Fandos et al. 2020) or might
476 have strong variation in their ecological niches during different parts of the year and thus vary in
477 niche-tracking. For example, individual white storks (*Ciconia ciconia*) and sandhill cranes
478 (*Antigone canadensis*) range from subtropical residents to short-distance migrants to longer-
479 distance migrants moving between northern temperate and tropical zones each year (Krapu et al.
480 2014; Fandos et al. 2020; Carlson et al. 2021). Summarizing niche tracking strategies at the

481 species level may limit our inferential ability in two ways: 1) species-level niche breadth is likely
482 greater than individual niche breadth given variability in individual niches (Carlson et al. 2021),
483 and 2) the accuracy of species-level niche centroids for individuals likely depends on the
484 variation in climate zones at which the species exists, so long as individuals and populations are
485 locally adapted (Araújo and Costa-Pereira 2013). Although most occurrence data is still available
486 at the species level, new technologies such as GPS tracking offer exciting opportunities for
487 assessing individual niches over time and across hierarchies of organismal organization (Jetz et
488 al. 2022; Costa-Pereira et al. 2022) and complement the presented species-level findings.

489
490 As climate change progresses, species are increasingly being exposed to conditions that are quite
491 different from those to which they have adapted (Pacifci et al. 2015). Gradual shifts in range
492 boundaries or phenology allow species to keep pace with long-term changes in climatic means
493 (La Sorte and Jetz 2012; Rushing et al. 2020; Koleček, Adamík, and Reif 2020). However, niche
494 tracking is an important tool for species to buffer both the changing mean environmental
495 conditions and increasing interannual variability associated with climate change by selecting
496 optimal environmental conditions over short time scales (Román-Palacios and Wiens 2020),
497 especially for species with narrow thermal or habitat requirements. Our findings reveal a broad
498 diversity in niche tracking strategies and uncover important functional trait associations with
499 migratory behavior and body size. This has ecological consequences for the assemblages and
500 ecosystems that will see different functional perturbations due to the different susceptibility of
501 niche tracking strategies to climatic change (Barbet-Massin and Jetz 2015). For example, as
502 niche tracking becomes more difficult with climate change, small-bodied and insectivorous birds
503 may be especially likely to fail to maintain their niches, resulting in proportional declines of
504 these species relative to the broader avian community assemblage.

505
506 Thus, we provide a framework for future studies to assess the degree of tracking both niche
507 position and breadth and capability for climate change adaptation in many other species globally,
508 including those from understudied regions with limited available data. However, additional work
509 is needed to assess how species have already been keeping up with climate change that has
510 occurred. Understanding how climate change is modifying species' ability to conserve their
511 niches will be a critical step towards determining which species are most likely to adapt to
512 increasing short-term variability in climate.

513

514

515 **Acknowledgements**

516

517 We thank J. Makinen and M. Lu for their modeling advice and thoughtful comments on the
518 analyses and J. Wilshire, A. Ranipeta, R. Li, and the Map of Life team at the Yale Center for
519 Biodiversity and Global Change for implementing the Spatiotemporal Observation Annotation
520 Tool to annotate occurrence data. We thank F. La Sorte for providing species-level migration

521 distances. We acknowledge funding from NSF grant DEB-1441737 and NASA grants
522 80NSSC17K0282 and 80NSSC18K0435.

523

524 **Data Sharing and Accessibility**

525

526 GBIF data is available for public use online at <https://www.gbif.org/>. Code generated to conduct
527 the analyses will be made available in a public repository such as Dryad or Figshare.

528

529 **References**

530

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- 697

Supporting Information

Title: Diverse strategies for tracking seasonal environmental niches at hemispheric scale

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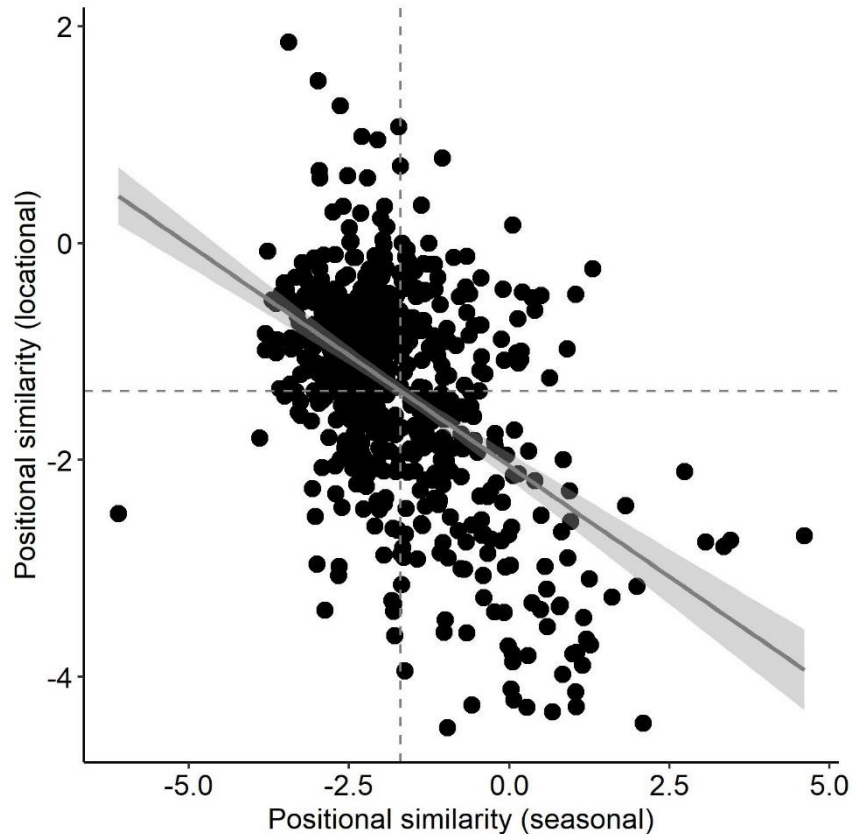


Figure S1. Seasonal and locational niche centroid similarity are inversely related. Niche-tracking species maximize niche similarity between breeding and nonbreeding seasons (defined as seasonal similarity, x-axis) and minimize the similarity between the nonbreeding season niche and the stationary winter niche (defined as locational similarity, y-axis). The reverse is true for non-tracker species. Tropical residents are an exception to this rule; these species have overlapping niche space for all three components, thus maximizing both seasonal and locational similarity. Dashed lines represent means and gray shading represents 95% confidence interval for trend line.

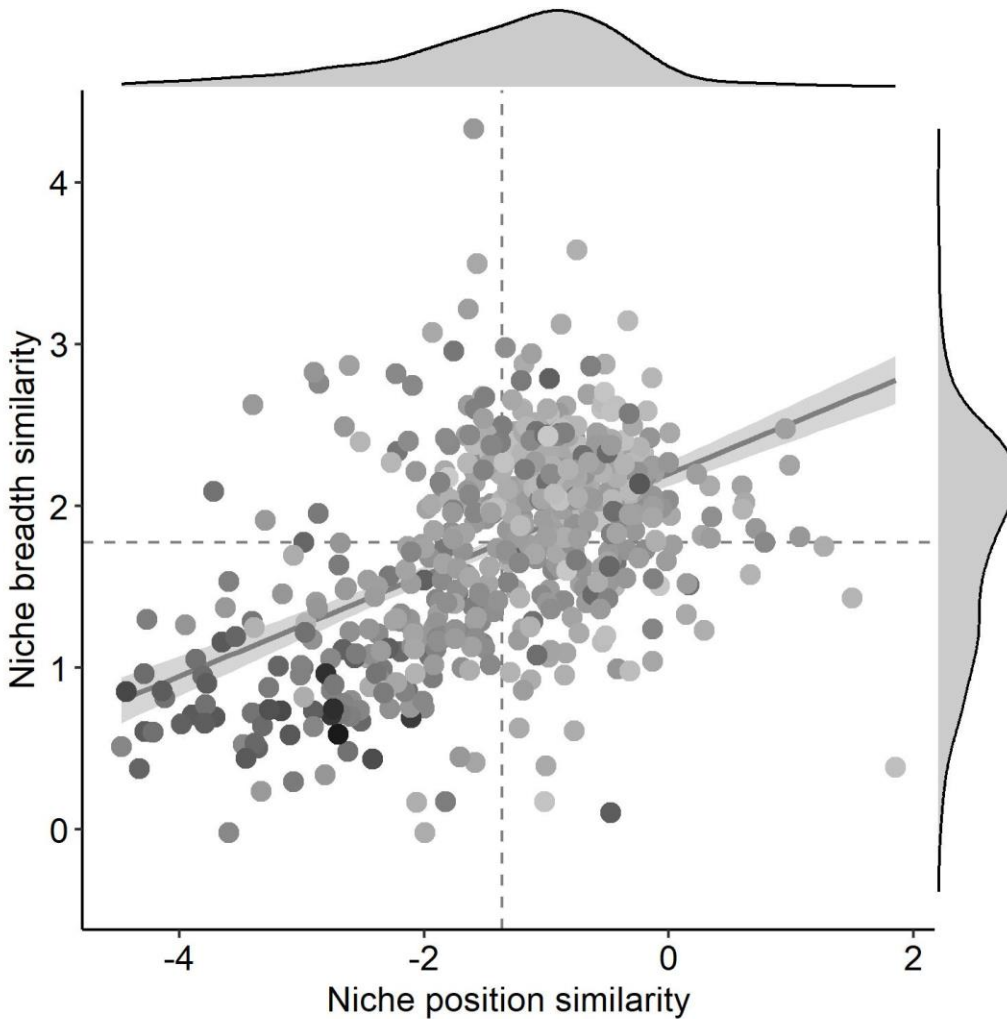


Figure S2: Interspecific variation in locational niche tracking of niche position and breadth.

Locational niche similarity represents the similarity of the overwintering niche to the winter conditions at breeding locations. Points represent species-level variation in locational similarity of niche position and breadth between the breeding and overwintering seasons for 619 species of North American birds. Marginal density plots illustrate a positive relationship between tracking locational niche position and breadth. Darker points reflect greater values of seasonal positional niche similarity and are primarily distributed among points with lower values of locational positional similarity.

Appendix 1. Seasonal niche dissimilarity for 619 species of North American birds (sorted taxonomically). Dissimilarity is partitioned into Mahalanobis distance and determinant ratio for both seasonal (breeding/overwintering comparison) and locational (overwintering/winter at breeding sites comparison) dissimilarity. Position and breadth similarity (visualized and used in analyses) was generated from these values by ln-transforming and sign-flipping.

(appendix_1.csv)

Appendix 2. Updates to avian phylogeny. We updated the avian phylogeny of Jetz et al (2012) to the 2021 Clements taxonomy by harmonizing species names, as follows.

(appendix_2.csv)

Table S1. Model tables (pgls and ANOVA) summarizing functional trait drivers of seasonal niche position and breadth similarity across species in multivariate models.

Mahalanobis distance				
PGLS table	Coefficient	SE	t-value	p-value
Intercept	-1.745	0.270	-6.472	<0.001
Migration distance	0.428	0.041	10.359	<0.001
Body mass (log)	-0.798	0.097	-8.233	<0.001
Diet (herbivore)	-0.647	0.178	-3.626	<0.001
Diet (insectivore)	-0.048	0.177	-0.275	0.784
Diet (omnivore)	-0.394	0.139	-2.836	0.005
Habitat (generalist)	0.001	0.174	0.005	0.996
Habitat (open)	-0.381	0.110	-3.470	0.001
Habitat (water)	0.364	0.172	2.110	0.035

Mahalanobis distance			
ANOVA table	DF	F-value	p-value
Intercept	1	121.713	<0.001
Migration distance	1	197.317	<0.001
Body mass (log)	1	93.094	<0.001
Diet	3	5.702	0.001
Habitat preference	3	9.675	<0.001

Determinant ratio				
PGLS table	Coefficient	SE	t-value	p-value
Intercept	1.360	0.262	5.183	<0.001
Migration distance	0.250	0.040	6.212	<0.001
Body mass (log)	0.525	0.094	5.565	<0.001
Diet (herbivore)	-0.053	0.174	-0.305	0.760
Diet (insectivore)	0.206	0.172	1.196	0.232
Diet (omnivore)	0.102	0.135	0.754	0.451
Habitat (generalist)	-0.081	0.169	-0.481	0.631
Habitat (open)	-0.400	0.107	-3.738	<0.001
Habitat (water)	0.319	0.168	1.905	0.057

Determinant ratio			
ANOVA table	DF	F-value	p-value
Intercept	1	185.261	<0.001
Migration distance	1	37.702	<0.001
Body mass (log)	1	51.669	<0.001
Diet	3	2.201	0.087

<u>Habitat preference</u>	3	10.145	<0.001
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Table S2. Model tables (pgls and ANOVA) summarizing functional trait drivers of locational niche position similarity across species in multivariate models.

Mahalanobis distance				
PGLS table	Coefficient	SE	t-value	p-value
Intercept	0.182	0.282	0.645	0.519
Migration distance	-0.331	0.043	-7.660	<0.001
Body mass (log)	0.952	0.101	9.389	<0.001
Diet (herbivore)	-0.215	0.187	-1.155	0.249
Diet (insectivore)	-0.176	0.185	-0.956	0.340
Diet (omnivore)	-0.092	0.145	-0.637	0.524
Habitat (generalist)	0.822	0.182	4.519	<0.001
Habitat (open)	0.415	0.115	3.609	<0.001
Habitat (water)	0.515	0.180	2.861	0.004

Mahalanobis distance			
ANOVA table	DF	F-value	p-value
Intercept	1	11.162	0.001
Migration distance	1	111.730	<0.001
Body mass (log)	1	111.154	<0.001
Diet	3	0.703	0.551
Habitat preference	3	7.757	<0.001

Table S3. Model tables (ppls) summarizing functional trait drivers of seasonal niche position and breadth similarity across species in univariate models.

Mahalonobis distance				
PGLS table	Coefficient	SE	t-value	p-value
Intercept	-3.060	0.214	-14.291	0.000
Migration distance	0.539	0.042	12.747	0.000

Mahalonobis distance				
PGLS table	Coefficient	SE	t-value	p-value
Intercept	-1.039	0.219	-4.744	0.000
Body mass (log)	-1.057	0.097	-10.883	0.000

Mahalonobis distance				
PGLS table	Coefficient	SE	t-value	p-value
(Intercept)	-1.254	0.257	-4.876	0.000
Diet (herbivore)	-1.287	0.202	-6.380	0.000
Diet (insectivore)	0.085	0.208	0.410	0.682
Diet (omnivore)	-0.681	0.155	-4.409	0.000

Mahalonobis distance				
PGLS table	Coefficient	SE	t-value	p-value
Intercept	-1.864	0.260	-7.168	0.000
Habitat (generalist)	-0.356	0.211	-1.684	0.093
Habitat (open)	-0.375	0.134	-2.796	0.005
Habitat (water)	0.121	0.208	0.578	0.563

Determinant ratio PGLS				
table	Coefficient	SE	t-value	p-value
Intercept	1.891	0.201	9.393	<0.001
Migration distance	0.229	0.040	5.770	<0.001

Determinant ratio PGLS				
table	Coefficient	SE	t-value	p-value
Intercept	1.938	0.201	9.659	<0.001
Body mass (log)	0.480	0.089	5.397	<0.001

Determinant ratio PGLS				
table	Coefficient	SE	t-value	p-value
(Intercept)	2.092	0.227	9.213	<0.001
Diet (herbivore)	0.071	0.178	0.401	0.688
Diet (insectivore)	0.354	0.183	1.934	0.054
Diet (omnivore)	0.435	0.136	3.190	0.001

Determinant ratio PGLS				
table	Coefficient	SE	t-value	p-value
Intercept	2.304	0.218	10.572	<0.001
Habitat (generalist)	-0.161	0.177	-0.909	0.364
Habitat (open)	-0.437	0.113	-3.881	<0.001
Habitat (water)	0.499	0.175	2.858	0.004

Table S4. Model tables (pgls) summarizing functional trait drivers of locational niche position similarity across species in univariate models.

Mahalonobis				
distance PGLS table	Coefficient	SE	t-value	p-value
Intercept	1.485	0.224	6.643	<0.001
Migration distance	-0.424	0.044	-9.610	<0.001

Mahalonobis				
distance PGLS table	Coefficient	SE	t-value	p-value
Intercept	-0.357	0.215	-1.658	0.098
Body mass (log)	1.119	0.095	11.721	<0.001

Mahalonobis				
distance PGLS table	Coefficient	SE	t-value	p-value
(Intercept)	0.331	0.265	1.252	0.211
Diet (herbivore)	0.521	0.208	2.510	0.012
Diet (insectivore)	-0.255	0.214	-1.194	0.233
Diet (omnivore)	0.351	0.159	2.207	0.028

Mahalonobis				
distance PGLS table	Coefficient	SE	t-value	p-value
Intercept	0.091	0.256	0.354	0.723
Habitat (generalist)	1.068	0.208	5.124	<0.001
Habitat (open)	0.408	0.132	3.085	0.002
Habitat (water)	0.774	0.205	3.770	<0.001

Table S5. Metrics describing phylogenetic signal in seasonal niche tracking. Asterisks denote significance level (***) < 0.001 < ** < 0.01 < * < 0.05).

	Blomberg's K		Lambda	
	Seasonal dissimilarity	Locational dissimilarity	Seasonal dissimilarity	Locational dissimilarity
Mahalanobis distance	0.083***	0.115***	0.664***	0.663***
Determinant Ratio	0.057*		0.278***	

