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Phylogenetic conservation of behavioral variation and behavioral syndromes

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Abstract

Individuals frequently differ consistently from one another in their average behaviors (i.e. 2 "animal personality") and in correlated suites of consistent behavioral responses (i.e. 3 4 "behavioral syndromes"). However, understanding the evolutionary basis of this 5 (co)variation has lagged behind demonstrations of its presence. This lag partially stems from comparative methods rarely being used in the field. Consequently, much of the 6 research on animal personality has relied on "adaptive stories" focused on single species 7 8 and populations. Here we used a comparative approach to examine the role of phylogeny in shaping patterns of average behaviors, behavioral variation, and behavioral correlations. In 9 comparing the behaviors and behavioral variation for five species of Gryllid crickets we 10 found that phylogeny shaped average behaviors and behavioral (co)variation. Variation in 11 average exploratory behavior and response to cues of predator presence attributable to 12 phylogeny was greater or comparable to the magnitude of "personality variation". 13 Likewise, magnitudes of variation were concordant with evolutionary relationships and 14 behavioral correlations were consistent across species. These results suggest that 15 phylogenetic constraints play an important role in the expression of animal personalities 16 17 and behavioral syndromes and emphasize the importance of examining evolutionary 18 explanations within a comparative framework.

19

Introduction

Behavioral syndromes, correlations between behaviors at the among-individual 20 level (Dingemanse et al. 2012), have been documented across taxa (Brommer and Class 21 2017). Behavioral syndromes can conceptually be thought of as correlations between 22 individual averages and stem from underlying genetic correlations and correlations due to 23 24 developmental plasticity and other sources of permanent environmental covariance (Dingemanse et al. 2012, Dingemanse and Dochtermann 2014). Among-individual variation 25 in behavior, often referred to as "personality variation", has been found to be similarly 26 ubiquitous (Bell et al. 2009). Similar to behavioral syndromes, this personality variation 27 can be thought of as variation across individuals in their average behaviors and likewise 28 29 stem from genetic and permanent environmental variation (Dingemanse and Dochtermann

2013, Dochtermann et al. 2015). Attempts to infer whether general taxonomic patterns 30 exist for both personality variation and behavioral syndromes have generally been 31 conducted via literature reviews and meta-analyses (Bell et al. 2009, Dochtermann 2011, 32 33 Garamszegi et al. 2012, 2013, Dochtermann et al. 2015, Brommer and Class 2017). These 34 synthesis efforts have shown that among-individual variation is common (average repeatability ~ 0.37 , Bell et al. 2009), that the magnitude of behavioral syndromes is 35 generally weak (average r \sim 0.19, Garamszegi et al 2012, 2013), and that there is general 36 alignment between nested patterns of correlations at the phenotypic, among-individual, 37 within-individual, and genetic levels (Dochtermann 2011, Brommer and Class 2017). 38

Despite the observation that both among-individual variation and behavioral 39 40 syndromes are common, we have a poor understanding of the evolution of either. This gap in our understanding is partly because comparative approaches have rarely been used in 41 studies of among-individual behavioral variation and behavioral syndromes (White et al. 42 2020), despite having been the backbone of studies of morphological evolution. Such 43 44 approaches allow for direct comparison across species of behavioral (co)variation and are necessary for a proper understanding of the importance of phylogeny in shaping 45 "personality" and behavioral syndromes (Rovauté et al. 2020, White et al. 2020). 46

Direct assessment of evolutionary hypotheses can also be extended to the study of
personality and behavioral syndromes: both among-individual variation and behavioral
syndromes have clear connections to quantitative genetic parameters; specifically, additive
genetic variation, and additive genetic covariances (Dochtermann and Roff 2010,
Dingemanse and Dochtermann 2014). The mathematical relationships between amongindividual (co)variances and additive genetic (co)variances (Boake 1989, Dingemanse and
Dochtermann 2014, Dochtermann et al. 2015) allows the extension of predictions from

54 quantitative genetics to among-individual variation and behavioral syndromes.

55 One such prediction is that differences in the magnitude of variation present for a 56 trait might be attributable to differences in selection between populations or species. 57 Specifically, Mousseau and Roff (1987) argued that traits with low heritability might be 58 indicative of strong selection having eroded genetic variation. Likewise, because among-59 individual variation represents the sum of additive genetic variation, dominance (and other 60 epistatic) genetic variation, and permanent environmental variation (e.g. irreversible and developmental plasticity), selection is expected to deplete this variation. Note, however,
that drift is often also expected to reduce genetic and, therefore, among-individual
variation.

64 Selection is likewise expected to shape additive genetic covariances and correlations, both by the loss of variation in single traits and changes to the magnitude and 65 directions of covariances (Roff 1997). For example, correlational selection is expected to 66 produce genetic correlations (Phillips and Arnold 1989, Armbruster and Schwaegerle 67 1996). As in the case of among-individual variances, these effects on genetic correlations 68 are expected to carry over to behavioral syndromes. In other words, behavioral syndromes 69 are expected to reflect the effects of selection on genetic correlations. Therefore, if 70 behavioral syndromes differ across populations, species, or other groupings, then this 71 72 suggests differences in genetic correlations and correlational selection (i.e. the "adaptive" hypothesis, Bell 2005). In contrast, if behavioral syndromes are conserved across groups 73 74 then this would suggest that either behavioral syndromes stem from pleiotropic effects (i.e. the "constraints" hypothesis, Bell 2005) or that selection is similar across groups. 75

While these topics have been addressed for other types of traits, particularly 76 morphological and chemical characteristics (Aguirre et al. 2014, Hine et al. 2014, 77 McGlothlin et al. 2018), addressing them for behavior remains important for several 78 reasons. First, considerable behavioral research assumes an adaptive framework for both 79 80 among-individual variation and behavioral syndromes, thereby minimizing the importance of phylogeny and minimizing the potential role of phylogenetic constraints. Second, 81 behaviors, life-history, and physiological traits exhibit substantially lower heritabilities 82 83 than do morphological traits (Mousseau and Roff 1987, Stirling et al. 2002, Dochtermann et al. 2019). Consequently, the role of phylogeny and selection in constraining and shaping 84 morphology may not generalize to traits with lower heritabilities and thus greater 85 plasticity. 86

Here, we compared the behavior of five closely related cricket species: *Gryllus integer, Gryllus assimilis, Gryllus lineaticeps, Gryllodes sigillatus,* and *Acheta domesticus*. For
each species we measured exploratory behavior and response to cues of predator
presence. By working with the same behavioral assays in five closely related species we

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were able to assess the importance of phylogeny for average behaviors and to evaluate 91 predictions about trait (co)variation. Specifically, we addressed the following questions: 92 1. Does the average expression of behavior differ among species? 93 94 We predicted that species would differ but do so in a manner constrained by 95 phylogeny. Put another way, more closely related species will have more similar average behaviors. 96 2. Do among-individual variances differ among species? 97 We did not have species level predictions but, because selection and drift should 98 both reduce among-individual variance, we predicted that among-individual 99 variation would differ across species independent of phylogeny. 100 3. Do within-individual variances differ among species? Within-individual variation, 101 typically disregarded as residual variation, includes phenotypic plasticity— 102 specifically reversible plasticity or "phenotypic flexibility" not captured by factors 103 and covariates of a statistical model (Piersma and Drent 2003, Whitman and 104 Agrawal 2009, Piersma and Van Gils 2011, Westneat et al. 2015, Berdal and 105 Dochtermann 2019). Differences across groups in the magnitude of within-106 individual variation therefore are, in part, differences in the magnitude of plasticity. 107 108 We did not have *a priori* expectations as to species differences or phylogenetic signal for within-individual variances. 109 4. Do behavioral syndromes differ among species? 110 Because behavioral syndrome structure has been conserved at the genetic level 111 across cricket populations of G. integer (Royauté et al. 2020), we predicted that 112 113 syndromes would similarly be phylogenetically conserved and shared across species. 114 Methods 115 Cricket Acquisition, Housing, and Rearing Conditions 116 Data used in this study were originally collected for various studies investigating the 117 effects of development on behavioral variation and the presence of behavioral constraints 118 and behavioral syndromes (Royauté et al. 2019, Royauté et al. 2020). A. domesticus males 119

and females were obtained as nymphs ($\sim 1 \text{ mm in size}$) from a commercial supplier 120 (Fluker's Cricket Farm, Port Allen, LA, U.S.A.) and were measured once mature. G. integer 121 females were captured in Aguila, AZ, G. lineaticeps males and females were caught in 122 123 Dunnigan, CA, and the *G. assimilis* males and females were caught in Maricopa County, AZ. These species were all captured during the summer of 2017. G. sigillatus individuals were 124 taken from an outbred population established by S. Sakaluk with crickets collected from 125 126 California and currently maintained in Fargo, ND. For *G. lineaticeps* and *G. assimilis*, the same individuals that were caught in the field were measured, while lab reared offspring of 127 *G. integer* were measured. All species were reared under a 12:12 light: dark photoperiod at 128 a temperature of 25-28°C. All individuals were housed in 0.71-liter containers with 129 130 transparent covers that included food, shelter, and water filled glass vials plugged with cotton balls. A. domesticus were exposed to a mixture of high and low quality diets 131 described in Royauté et al. (2019), while all other species included in this study were fed ad 132 libitum food (commercially purchased chicken feed). 133

134 Behavior Trials

To measure exploratory behavior and anti-predator responses we repeatedly 135 recorded individuals' activity levels in an open field arena, followed by their responses to 136 cues of predator presence created from diluted *Eublepharis macularius* excreta (see details 137 below). A. domesticus were measured between March 2015 and October 2016, G. lineaticeps 138 were measured from August 2017 to September 2017, *G. assimilis* were measured between 139 September 2017 and October 2017, *G. integer* were measured between May 2018 and June 140 2018, and *G. sigillatus* were measured in May 2019. All trials were conducted in a plastic 141 arena (60 cm x 60 cm and 15 cm high) with a Plexiglas lid. The arena was split into four 30 142 cm × 30 cm arenas separated by a divider, allowing up to four crickets to be tested at one 143 time. Open field trials were always conducted first followed by antipredator response trials 144 145 either immediately after or on another day to minimized potential carryover effects from exposure to cues of predator presence. After each behavioral assay, arenas were 146 thoroughly cleaned with 70% ethanol wipes to avoid accumulation of any chemical traces 147 of conspecifics. Mass at the time of behavioral trials was recorded to the nearest 1 mg. All 148 individuals were measured in each assay for a maximum of three repetitions, with some 149

- individuals measured fewer times due to escape or natural mortality (Table 1). In total, we
- 151 conducted 2478 behavioral assays across a total of 460 individuals (Table 1).
- 152

Species	Behavioral Assay	Repetition 1	Repetition 2	Repetition 3	Total Trials
Achata domastique	Open field	281	263	225	769
Acheta aomesticus	Antipredator	262	235	220	717
Cmillus assimilis	Open field	16	16	16	48
Grynus assimilis	Antipredator	16	16	15	47
Cmillus integer	Open field	92	91	74	257
Grynus mieger	Antipredator	88	88	72	248
Cmillus lineaticons	Open field	21	17	11	49
Grynus inteuticeps	Antipredator	21	13	11	45
	Open field	50	50	49	149
Grynoaes signatus	Antipredator	50	50	49	149
Total		896	837	743	2478

Table 1. Number of individuals, by species, for which behavior was assayed in a first, second, and third repetition.

153 *Open field behavior*

154 Individual crickets were left to rest for 30 seconds under a 5 cm diameter cup after being introduced into the lower right section of the arena (Figure S1). After these 30 155 seconds we allowed the individuals to move freely through the arena for 220 seconds. We 156 157 measured each individual's exploratory propensity by calculating the number of *unique* zones visited (UZ) by the cricket with Ethovision X (Noldus Information Technology, 158 Wageningen, The Netherlands). This behavioral protocol has previously been used with *A*. 159 *domesticus* and *G. integer* to evaluate genetic and individual differences in activity and 160 exploratory behaviors (Royauté et al. 2015, Royauté and Dochtermann 2017, Royauté et al. 161 2019, Royauté et al. 2020). 162

163 *Predator cue response*

To measure responses to cues of potential predator presence, we collected excreta from three adult leopard geckos, *Eublepharis macularius*, that were fed a mixed diet of *A*. *domesticus, G. sigillatus, G. lineaticeps, G. integer*, and *G. assimilus*. Leopard geckos were housed according to the standards of the Institutional Animal Care and Use Committee of

North Dakota State University (Protocol A14006, A17015, and A19067) and the Animal 168 Behavior Society (2020). Collected excreta was frozen and then finely ground and diluted 169 with deionized water (1 ml H2O: 5 mg of excreta). This solution was then applied to 15 cm 170 171 diameter filter paper disks with a 5 cm diameter central cutout that allows crickets to be left to rest unexposed to the predator cues (Royauté and Dochtermann 2017, Royauté et al. 172 2019, Royauté et al. 2020). Each predator cue disk was left to dry for a minimum of 2 hours 173 174 then stored at -23°C until needed for trials. Predator cue disks were allowed to warm to room temperature before use in antipredator trials and discarded after a single use. 175 Between each trial, cue disks were stored at 4°C for a maximum of 14 days. 176

177 We placed the predator cue disk at the bottom of a 15 cm diameter arena and left the cricket to rest for a minimum of 30 seconds under a 5 cm diameter cup in the 178 179 nontreated central cutout. We then removed the cup and allowed the cricket to move freely for 220 seconds and estimated the distance travelled in cm (AP distance) using Ethovision 180 X (Figure S1). Previous studies with this protocol show that crickets had heightened 181 182 activity levels in the presence of this diluted gecko excreta compared to water controls (Royauté and Dochtermann 2017). Consistent with this, G. sigillatus crickets have been 183 184 found to increase their activity after direct exposure to predators (Bucklaew and 185 Dochtermann 2020). Greater activity during these antipredator response assays, i.e. 186 greater AP distance, was therefore interpreted as a greater responsiveness to predator 187 cues.

188 Data Analysis: Univariate Models

To assess differences in behavioral responses between species for means and 189 variances we analyzed behavioral data using separate univariate mixed-effects models for 190 unique zones visited and AP distance (square root transformed). We included species, 191 temperature (Celsius, mean centered), mass (using among- and within-individual centering 192 193 (Van de Pol and Wright 2009)), and sex as fixed effects. Individual ID was included as a random effect. We compared the fit of four univariate mixed models structured as follows: 194 1) Model 1: $V_i = \& V_w = A$ null model where the among- (V_i) and within-individual (V_w) 195 variances were kept constant between species. 196

2) Model 2: $V_i \neq \& V_w = A$ model where the among-individual variance differed between 197 species, but the within-individual variance was kept constant. 198

3) Model 3: $V_i = \& V_w \neq$ The within-individual variance differs between species, but the 199 200 among-individual variance was kept constant.

201

4) Model 4: $V_i \neq \& V_w \neq$ Both the among and within-individual variances were allowed 202 to vary between species.

203 These models were specified using the MCMCglmm package for Bayesian mixed models (Hadfield 2010) using Markov-chain Monte Carlo (MCMC) with 1.3 million iterations, 204

300,000 iteration burn-in, a thinning interval of 1000, and an inverse-Wishart prior. AP 205

206 distance and unique zone models were fit with Gaussian and Poisson error distributions,

respectively. 207

208 To determine whether species differed in average behavior, Models 1 and 4 were run 209 with and without species as a fixed effect and compared based on deviance information criterion (DIC) values. If species differ in average behavior, models with species included as 210 a fixed effect would be expected to have lower (DIC) values. Average behavioral differences 211 among species reported in the Results section were then qualitatively assessed using 212 posterior-modal estimates for each species (Congdon 2006). 213

214 We then compared DIC values among models 1 through 4 to determine whether either among- or within-individual variances differed among species following Royauté et al. 215 (2019) and Royauté and Dochtermann (2020). The model with the lowest DIC value was 216 217 considered the best model and models with Δ DIC>5 were considered to have a substantively poorer fit (Barnett et al. 2010). Models with Δ DIC<5 were considered as 218 219 having comparable support relative to the best model (Barnett et al. 2010). All models 220 were specified with the same fixed effect structure as specified above to prevent biased estimates of variance components and repeatability (Spiegelhalter et al. 2003, Nakagawa 221 and Schielzeth 2010b, Westneat et al. 2011). 222

223 Data Analysis: Phylogenetic Signal

As our primary questions were about differences in behavioral averages and 224 variances, our results and discussion focus on the above model comparisons. However, we 225 also calculated the variation in behavior directly attributable to phylogeny. To do so, we fit 226

mixed effects models with the same fixed effects, prior structure, and chain parameters as 227 above but omitting Species as a fixed effect. Species was instead incorporated as a random 228 effect, along with individual ID, with the relationship among species modeled according to 229 230 the current phylogeny (Figure 1, Weissman and Gray (2019)). From these models we then 231 estimated the strength of phylogenetic signal as the proportion of variation attributable to the hierarchical pattern of relatedness among species (i.e. λ , Pagel 1999, Hadfield and 232 233 Nakagawa 2010, Nakagawa and Santos 2012). From the same models we also estimated the proportion of variation attributable to among-individual differences (i.e. τ, repetability, 234 Dingemanse and Dochtermann 2013). We estimated both phylogenetic signal and 235 repeatability as unadjusted values; that is, we included the variation attributable to fixed 236 237 effects in the ratio denominator (Nakagawa and Schielzeth 2010a).

238 Data Analysis: Bivariate Models

Behavioral syndromes were estimated using bivariate mixed-effects models with 239 240 unique zones traveled and AP distance as response variables, also using the MCMCglmm 241 library (Hadfield 2010), and analyzed separately for each individual species. We fit models using temperature (Celsius, mean centered), mass (using among- and within-individual 242 centering on subjects (Van de Pol and Wright 2009)), and sex as fixed effects and individual 243 ID was fit as a random effect. These models were fit with 2.6 million iterations, a 600,000 244 burn-in period, a thinning interval of 2000, and a prior that was flat for correlations. 245 Among-individual correlations were estimated for all species, while within-individual 246 correlations were only assessed when individuals were measured for unique zones 247 traveled and antipredator activity during the same testing period (Dingemanse and 248 249 Dochtermann 2013). Consequently, we were unable to assess within-individual covariation of *G. lineaticeps* and *G. assimilis* due to the fact that these species were not measured for 250 each behavior in immediate succession. Because model comparisons as used above for 251 single traits could not be conducted for correlations (due to software imposed model 252 limitations), differences in behavioral correlations across species were assessed based on 253 whether 95% HPD intervals overlapped. Overlap of 95% intervals is an over-conservative 254 comparison metric (Royauté and Dochtermann 2020), but this did not affect our species 255 comparison results here. All analyses were conducted in R 3.4.4 (Team 2018). 256

257

Results

258 Differences in average behavior among species

Species differed in average behaviors: the inclusion of species as a fixed effect substantially improved model fit for both behaviors (Table 2, Table S1). The monophyletic group of *G. assimilis, G. integer*, and *G. lineaticeps* exhibited the lowest number of unique zones visited (Figure 1a) but differences in AP distance were less obviously associated with phylogenetic structure (Figure 1b). Consistent with this, phylogenetic signal was stronger for unique zones visited ($\lambda : 0.27$) than for AP distance ($\lambda : 0.16$; Table S2).

Table 2. DIC values for statistical models with and without the inclusion of species as a fixed effect. The effect of species was evaluated in a model where variances did not (Model 1) or did (Model 4) differ by species. For both behaviors and both models, the inclusion of species substantially improved model fit, as indicated by the lower DIC values for models with species included as a fixed effect.

	Behavior	DIC with species	DIC without species	DIC(without) – DIC(with)
Model 1	AP Distance	8025.51	8058.71	32.2
$(V_i = \& V_w =)$	Unique Zones Visited	8982.97	8456.17	526.8
Model 4	AP Distance	7763.82	7780.94	17.12
$(V_i \neq \& V_w \neq)$	Unique Zones Visited	8338.21	8344.88	6.67



Figure 1. Species posterior-modal values with 95% HPD credibility intervals. (a) Average unique zones visited. (b) Average AP distance in centimeters. (c) Among-individual variances in unique zones traveled. (d) Among-individual variances in AP distance. (e) Within-individual variances in unique zones traveled. (f) Within-individual variances in AP distance.

265 Differences in variances among species

The best fit model for unique zones visited was Model 4, which allowed both among
and within-individual variances to vary across species. All other models were poorly
supported (ΔDIC>8; Table 3). This indicates that both among- and within-individual

variances differed among species in open field trials. For AP distance, Models 3 and 4 fit

270 comparably well (Table 3). Both of these models support differences among species in

- 271 within-individual variances for AP distance. The difference between the models therefore
- 272 suggests mixed support for species differences in among-individual variances for AP
- 273 distance.

The monophyletic group of *G. assimilis, G. integer,* and *G. lineaticeps* exhibited higher among-individual variation for unique zones visited and lower among-individual variation for AP distance (Figure 1c & d). This monophyletic group also exhibited higher withinindividual variation for both unique zones visited and AP distance than observed for *A. domesticus* and *G. sigillatus* (Figure 1e & f).

Model (variance constraints)	Behavior	DIC	ΔDIC
Model 1 (V_i = & V_w =)	Unique Zones Visited	8982.97	644.76
Model 2 ($V_i \neq \& V_w =$)	Unique Zones Visited	8420.69	82.48
Model 3 ($V_i = \& V_w \neq$)	Unique Zones Visited	8346.44	8.23
Model 4 (V _i ≠ & V _w ≠)	Unique Zones Visited	8338.21	0
Model 4 ($V_i \neq \& V_w \neq$) Model 1 ($V_i = \& V_w =$)	Unique Zones Visited AP Distance	8338.21 8025.51	0 263.31
Model 4 ($V_i \neq \& V_w \neq$) Model 1 ($V_i = \& V_w =$) Model 2 ($V_i \neq \& V_w =$)	Unique Zones Visited AP Distance AP Distance	8338.21 8025.51 8010.04	0 263.31 247.84
Model 4 ($V_i \neq \& V_w \neq$) Model 1 ($V_i = \& V_w =$) Model 2 ($V_i \neq \& V_w =$) Model 3 ($V_i = \& V_w \neq$)	Unique Zones Visited AP Distance AP Distance AP Distance	8338.21 8025.51 8010.04 7762.20	0 263.31 247.84 0

Table 3. DIC and Δ DIC values of model fit for AP distance and unique zones visited.

279 Differences in behavioral correlations among species

Among-individual behavioral correlations were of similar magnitude for *A*. *domesticus, G. assimilis, G. lineaticeps*, and *G. sigillatus* (0.3 : 0.5, Figure 2a, Table S3) while
the correlation for *G. integer* was estimated to be slightly higher (0.66, Figure 2a, Table S3).
Importantly, the lower bounds of the HPD intervals for *G. assimilis, G. integer, G. lineaticeps*,
and *G. sigillatus* also overlapped with 0 (Figure 2, Table S3). This is perhaps unsurprising
given the small sample sizes for *G. assimilis* and *G. lineaticeps*.

Behavioral correlations at the within-individual level ranged from 0.1 to 0.35 for *A. domesticus, G. sigillatus,* and *G. integer*, with *G. integer* having the lower bound of its HPD interval overlapping with 0 (Figure 2b). The overlapping of 0 indicates that behavioral plasticity might not be integrated in this species. Behavioral correlations at either level did not show obvious patterns relative to phylogeny and were not significantly different across species (Figure 2).



Figure 2. Species posterior-modal values with 95% HPD credibility intervals. (a) Amongindividual behavioral correlations of unique zones visited and AP distance. (b) Withinindividual differences of behavioral correlations of unique zones traveled and AP distance. Within-individual correlations for *G. assimilis* and *G. lineaticeps* were not calculated as behavior trials were not performed in close succession.

292

Discussion

- 293 Our results demonstrate that species differed in their exploratory behavior and
- 294 response to cues of predator presence at all levels of variation but that behavioral
- 295 syndromes were conserved across species. These results suggest an important influence of
- 296 phylogenetic constraints on how behaviors evolve.

297 Species differed from one another in their average behaviors (Table 2), in a manner consistent with phylogenetic relationships. Specifically, the monophyletic group of G. 298 299 assimilis, G. integer, and G. lineaticeps were generally similar in average unique zones 300 visited (Figure 1a). In contrast, while average AP distance differed by species, it did not do 301 so in a manner clearly concordant with phylogeny (Figure 1b). Indeed, phylogenetic signal, the proportion of variation attributable to the hierarchical pattern of relatedness among 302 303 species, was higher for unique zones visited than for AP distance (Table S2). Interestingly, 304 and relevant for future research, phylogeny explained considerably more variation in our measure of exploratory behavior—unique zones visited—than did among-individual 305 306 variation, i.e. "animal personality" ($\tau = 0.15$ versus $\lambda = 0.27$; Table S2).

307 The species we examined also differed in among-individual variation in exploratory (unique zones visited) and predator response (AP distance) behaviors, again in a manner 308 consistent with phylogenetic relationships (Figures 1c, d). Unfortunately, phylogenetic 309 310 methods have been developed primarily with the goal of understanding differences in trait averages rather than trait (co)variances. Our comparisons of "personality" variation and 311 syndromes among species are therefore based on the model comparison methods 312 313 identifying the presence of species differences and subsequent qualitative comparisons of 314 species level estimates. Nonetheless, the concordance between patterns of the magnitude of among-individual variation and the currently described phylogeny suggests 315 phylogenetic constraints on the magnitude of "personality" variation. Of the five species, 316 317 the monophyly of *G. assimilis, G. integer*, and *G. lineaticeps* exhibited the highest among-318 individual variation in unique zones visited and the lowest among-individual variation in 319 AP distance (Figure 1c and d). While the expression of average behaviors and behavioral syndromes might be expected to exhibit phylogenetic signal, we did not expect this to be 320 the case for among-individual variances. One possible explanation would be bottlenecks at 321 more basal phylogenetic nodes leading to reduced genetic variation present in subsequent 322 groupings. While this could explain the lower among-individual variation in AP distance for 323 324 G. assimilis, G. integer, and G. lineaticeps, it does not explain that those same species exhibit 325 higher within-individual variation for unique zones visited.

Differences observed in among-individual variation could also be attributable to selection differentially acting upon these species by reducing the additive genetic variation present in a population or species (Mousseau and Roff 1987). Our results therefore suggest the possibility that exploratory behavior, for which unique zones visited is a proxy, has been under stronger selection for *A. domesticus* and *G. sigillatus* than for the other species. Importantly, because we do not know the strength and direction of selection acting on these phenotypes, the data presented here only suggests this possibility and cannot be used to distinguish between the effects of selection and drift for either behavior.

334 Alternative explanations for the observed differences in among-individual variances stem from differences across source populations and sampling of these populations. For 335 example, individual A. domesticus used in this study were from a captive population where 336 inbreeding could have reduced genetic variation over generations. This potentially explains 337 338 the low among-individual variation the species shows for unique zones visited (Figure 1c) 339 but is contradicted by the high among-individual variation in AP distance (Figure 1d). In contrast, *G. assimilis* and *G. lineaticeps* behavior was measured for field-caught individuals. 340 341 If individuals of these species experienced different developmental environments from one 342 another, we would predict higher among-individual variation in behavior because permanent environmental variation contributes, on average, 50% of the observed among-343 individual variation in behavior present in populations (Dochtermann et al. 2015). This 344 explanation is not, however, supported: while G. assimilis and G. lineaticeps indeed showed 345 high relative among-individual variation in unique zones visited, the same was not the case 346 347 for AP distance (Figure 1c and d). Moreover, for both behaviors, G. assimilis and G. *lineaticeps* were very similar to *G. integer*, for which lab reared individuals were measured. 348 349 To summarize, the conflicting patterns of among-individual variation observed between AP 350 distance and unique zones visited prevents clear interpretation.

Estimated within-individual variances include variation from a variety of sources, 351 352 including plasticity in response to short-term environmental variation and measurement error (Dingemanse et al. 2012, Berdal and Dochtermann 2019). In comparing species, if we 353 assume measurement error is similar among species, differences in within-individual 354 355 variation will primarily represent differences in plasticity. This short-term plasticity, also 356 referred to as phenotypic flexibility (Piersma and Van Gils 2011), allows individuals to 357 respond flexibly to an environment (Westneat et al. 2015). As was the case for among-358 individual variation and average unique zones visited, G. assimilis, G. integer, and G.

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lineaticeps were grouped together and exhibited similar magnitudes of within-individual
variation (Figure 1e and f). For both behaviors, this group exhibited considerably higher
within-individual variation than observed for *A. domesticus* and *G. sigillatus*, differences
supported by our model comparison results (Table 3). In other words, the *Gryllus* genus
exhibited greater behavioral plasticity.

One possible explanation for this pattern is that our sample of *G. assimilis*, *G. integer*, 364 365 and G. lineaticeps were of individuals either caught from the field or the direct offspring of 366 field inseminated and subsequently captured individuals. In contrast, the population of G. sigillatus we sampled had been in captivity for around 75 generations and the population of 367 A. domesticus was reared for production purposes for some undetermined but large 368 number of generations. Consequently the differences in within-individual variation could 369 370 be attributable to exposure to a frequently changing environment (Relyea 2001) in the case of G. assimilis, G. integer, and G. lineaticeps and the loss of plasticity in A. domesticus and G. 371 372 *sigillatus.* This possibility could be assessed for crickets via experimental evolution with 373 populations experiencing different levels of environmental heterogeneity.

With regard to behavioral correlations, Bell (2005) proposed two hypotheses for the 374 375 expression of behavioral syndromes within a population relevant to the species level comparisons we performed. The first of these, the constraints hypothesis, chiefly attributes 376 behavioral syndromes to the presence of pleiotropy, with the expression of genes affecting 377 378 multiple behaviors. This hypothesis can be extended to other mechanistic connections 379 constraining independent trait expression. Second, the adaptive hypothesis states that behavioral syndromes are the adaptive outcome of correlated selection. While pleiotropy 380 381 and other mechanistic connections can evolve and be adaptive, syndromes attributable to 382 the adaptive hypothesis are expected to respond more quickly to changes in selection (Roff 1997). Consequently, phylogenetic similarity in behavioral syndromes provides indirect 383 support for the constraints hypothesis. Due to among-individual correlations not 384 substantively differing among species (Figure 2), our results therefore support the 385 386 constraints hypothesis, despite species differing in variances and average expressions of 387 behaviors (Figure 1).

While a comparative approach has only rarely been used for examining behavioral variation, three particular studies are relevant to the interpretation of our results here. 390 First, Blankers et al. (2017) compared the phenotypic variances and (co)variances of seven calling traits of multiple cricket species (including *G. lineaticeps*, which was included in our 391 study). These authors found that the phenotypic covariance matrices differed among 392 393 cricket species. One of the major differences among species was in the magnitude of 394 variation present in single traits (Blankers et al. 2017). This is consistent with our findings that variances of behaviors differed across species (Figure 1c-f). Unfortunately, these 395 396 authors compared phenotypic (co)variances, which conflate among- and within-individual 397 (co)variation (Dingemanse et al. 2012). Second, White et al. (2020) compared the amongindividual covariance matrices of seven species of fish. Comparable to our results, these 398 399 authors detected differences in the magnitude of among-individual behavioral variability 400 and also found overall phylogenetic signal and similarity in how variation was expressed 401 across multiple behaviors (White et al. 2020). Finally, Royauté et al. (2020) compared the 402 expression of additive genetic (co)variance (i.e. **G** matrices) in behavior among four populations of *G. integer*. Similar to White et al. (2020) and our results presented here, 403 Royauté et al. (2020) found differences in single trait variances and covariances but the 404 overall structure of trait covariance was generally conserved across populations— 405 indicating support for the constraints hypothesis. 406

More generally, our findings here suggest that behavioral correlations are
phylogenetically conserved. Conserved trait correlations like those observed here
constrain the divergence of populations and species (Schluter 1996). While the potential
for such constraints has been speculated about for behaviors (Dochtermann and
Dingemanse 2013), prior demonstrations of such have primarily focused on morphological
traits (McGlothlin et al. 2018, Sztepanacz and Houle 2019) and chemical traits (Blows et al.
2004, Aguirre et al. 2014).

Jointly, our approach allowed us to determine whether there were differences in average behavior, "personality", behavioral plasticity, and behavioral syndromes among species. Our results demonstrate phylogenetic conservation of behavioral averages, behavioral variation, and behavioral syndromes. This finding is potentially surprising given that behavior is often assumed to be more flexible and labile than other types of traits (but see Zuk and Spencer 2020) and suggests an important role for phylogenetic constraints as

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Supplemental Materials

Table S1. Fixed effects coefficients for Model 4 (Table 3). The intercept estimate is for *Acheta domesticus* females (fixed effect coefficients are contrasts versus these values).

	posterior	95% credib	ility interval	effective	nMCMC
	mean	lower	upper	sample size	рмсмс
Intercept	16.97	16.25	17.66	1000	< 0.001
Gryllus assimilis	-1.44	-4.18	1.22	1000	0.298
Gryllus integer	3.67	2.31	5.18	1000	< 0.001
Gryllus lineaticeps	-0.52	-4.00	2.95	1000	0.782
Gryllodes sigillatus	7.43	5.69	9.17	1000	< 0.001
Temp2	0.65	0.34	0.96	1000	< 0.001
SexM	-0.40	-1.44	0.72	1000	0.454
Mass (w/in individual centered)	0.72	-0.77	2.29	1000	0.344
Mass (b/w individual centered)	0.02	-0.76	0.98	800.8	0.972
Unique zones visited					

AP Distance (square-root transformed)

	posterior	95% credib	ility interval	effective	pMCMC
	mean	lower	upper	sample size	pmcmc
Intercept	2.929	2.869	2.984	1000	< 0.001
Gryllus assimilis	-1.227	-1.757	-0.628	1000	< 0.001
Gryllus integer	-0.728	-0.891	-0.526	1000	< 0.001
Gryllus lineaticeps	-0.593	-0.979	-0.219	1000	0.004
Gryllodes sigillatus	0.282	0.160	0.411	1000	< 0.001
Temp2	0.043	0.016	0.070	815.1	0.002
SexM	0.111	0.010	0.201	1000	0.022
Mass (w/in individual centered)	-0.022	-0.175	0.135	1000	0.792
Mass (b/w individual centered)	0.120	0.043	0.212	899.8	0.006

Table S2. Variance estimates (posterior modes with 95% credibility intervals) for models including phylogenetic structure as a random effect. Models were fit with temperature (centered), sex, and mass (within and between individual centered) as fixed effects. Phylogeny was modeled according to the trees shown in Figure 1 and with uniform branch lengths. Subject was also included as a random effect. Variance ratios are presented as unadjusted ratios; that is, variance due to fixed effects is included in the denominator. Ratios for unique zones include the distribution specific variance (DSV) in the denominator. λ and τ correspond to unadjusted phylogenetic signal and unadjusted repeatabilities respectively.

	Variance estimate	Variance ratios*
	(95% CrI)	(95% CrI)
AP Distance		
Phylogeny	12.73 (2.10 : 106.61)	λ : 0.16 (0.05 : 0.68)
Subject	13.34 (10.21 : 17.66)	τ: 0.19 (0.07: 0.27)
Fixed Effects	1.10 (0.32 : 2.60)	0.01 (0:0.04)
Residual	38.34 (33.96 : 41.50)	0.50 (0.24 : 0.71)
Unique Zones		
Visited		
Phylogeny	0.12 (0.03 : 1.07)	$\lambda : 0.27 (0.09 : 0.72)$
Subject	0.11 (0.07 : 0.13)	τ: 0.15 (0.05: 0.23)
Fixed Effects	0.01 (0:0.02)	0.01 (0:0.03)
Residual	0.26 (0.23 : 0.31)	0.48 (0.16 : 0.58)
DSV**	0.06 (0.03 : 0.11)	NA

* while the ratios for any single MCMC estimate will sum to 1, the posterior modes can sum to other values due to uncertainty across the MCMC chain

** estimated as $ln\left(\frac{1}{exp(\beta_0)}+1\right)$ following Nakagawa & Schielzeth (2010)

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Species	Among-individual correlation (95% CrI)	Within-individual correlation (95% CrI)
Gryllus assimilis	0.37 (0.21 : 0.54)	NA
Gryllus integer	0.29 (-0.63 : 0.75)	0.10 (-0.08 : 0.25)
Gryllus lineaticeps	0.66 (-0.01 : 0.82)	NA
Gryllodes sigillatus	0.49 (-0.42 : 0.78)	0.35 (0.12 : 0.53)
Acheta domesticus	0.3 (-0.05 : 0.6)	0.18 (0.08 : 0.27)

Table S3. Among- and within-individual correlations by species. Correlation estimates are posterior modes and are presented along with 95% credibility intervals (CrI).



Figure S1. Schematics of the open field (left) and predator cue arenas (right). The open field arenas were subdivided into 36 unique "zones" during video processing. For the antipredator response trials the cricket was introduced, under a container, to the center point. This cricket release zone did not have predator cues present.