A Phylogenetic Analysis of Shape Covariance Structure in the Anthropoid Skull

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Abstract

Phenotypic traits evolve in a coordinated manner due to developmental and functional interactions, mediated by the dynamics of natural selection; the dependence between traits arising from these three factors is captured by genetic (G) and phenotypic (P) covariance matrices. Mammalian skull development produces an intricate pattern of tissue organization and mutual signaling that integrates this structure, although the set of functions it performs is quite disparate. Therefore, the interplay between these interactions, and their relationships with the adaptive landscape may thus influence divergence in covariance structure among sister lineages. Here, we evaluate the stability of phenotypic covariance structure in skull size and shape along the diversification of Anthropoid Primates under a explicit phylogenetic framework. We estimate 10 diversity in covariance structure, testing hypotheses concerning the phylogenetic distribution 11 of P-matrix variation and pinpoint which traits are associated with this variation. We find that most changes occurred in the basal split between Platyrrhini and Catarrhini, and that these changes occurred within both Orbital and Basicranial trait sets, while Oral, Nasal and Vault trait sets present stable associations along the Anthropoid phylogeny. Therefore, changes in P-matrix structure among Anthropoids are restricted to trait sets whose functional significance is associated with the accommodation of the two precursor tissues that compose the skull, while the stability in the remaining regions hints at the stability of the underlying functional relationships imposed by the adaptive landscape.

Introduction

Phenotypic traits evolve in a coordinated manner either because of shared genetic and developmental processes or joint effects on fitness (Gould & Lewontin, 1979; Lande, 1979; Felsenstein, 1988; Zeng, 1988), and response to natural selection is maximized when both factors (variation and selection) are aligned, while discordance between them may deflect evolutionary response away from maximum increase in fitness (Schluter, 1996). The additive genetic covariances among traits (G) and the partial regression coefficients of fitness on traits (β) provide linear approximations for both these effects in the characterization of phenotypic change across generations (Lande, 1979; Rice, 2002). Therefore, the additive covariances in G represent the 28 codependency between traits due to pleiotropy and linkage disequilibrium, characterizing a 29 linear genotype/phenotype map centered on the mean phenotype (Wagner, 1984, 1996; Cheverud, 1996a). 31 However, the structure of pleiotropic interactions depends on the local curvature of the genotype/phenotype map, traditionally represented in quantitative genetics as either dominance or epistasis (Rice, 1998, 2004; Wolf et al., 2001). The effect of epistatic loci on the modulation of pleiotropic interactions has been identified in experimental settings (e.g. Cheverud et al., 2004; Wolf et al., 2005; Pavlicev et al., 2008), indicating that populations may harbor genetic variation in the association between traits. Genetic covariances among phenotypic traits then evolve as a consequence of changes in allele frequencies in these loci, for example in response to genetic drift and founder effects (Goodnight, 2000; Brito et al., 2005). Local features of the adaptive landscape may also impact genetic covariances among traits, as G is thought to match the patterns imposed by stabilizing selection and mutational effects (Lande, 1980; Cheverud, 1984; Jones, 2007). It is noteworthy that changes in genetic covariances due to the curvature of adaptive landscapes can be explained just by considering shifts in linkage disequilibrium among loci (Turelli, 1988), without need to appeal to epistatic pleiotropy. However, recent experimental data (Careau et al., 2015) and simulation-based models (Jones et al., 2012; Melo & Marroig, 2015)

have demonstrated the effects of directional selection on the structure described by G, thus indicating that the linear component of adaptive landscape can also have an impact on genetic covariances due to genetic variation in pleiotropic interactions. The relationship between adaptive landscapes and intrapopulational covariance structure is 49 mediated through performance (Arnold, 1983), which may be thought as a dynamical property of 50 phenotypes. This implies that the separation between developmental and functional interactions 51 as two distinct factors shaping phenotypic covariance structure is blurry at best (Cheverud, 1996a; 52 Zelditch & Swiderski, 2011). For example, in the mammalian skull, precursor tissues originate two 53 distinct regions, Face and Neurocranium, which exhibit marked contrasts in terms of functions 54 which they perform, interactions with soft tissues, and response to developmental milestones, 55 such as birth or weaning (Hallgrímsson & Lieberman, 2008; Lieberman, 2011). Thus, the Oral, Nasal, and Zygomatic regions are associated with the Face and are responsible for mastication, respiration, and the attachment of the muscular apparatus involved in mandibular articulation. The Vault, Orbit and Base regions are associated with the Neurocranium and are responsible for encasing and protecting both brain and eye, and for supporting and connecting the skull with the rest of the body. The Vault also houses muscle attachment sites associated with mandibular articulation, indicating that, to some degree, regions may be involved in more than one function 62 and that some functions may be shared between them. The contrast between these regions is the 63 result of distinct gene expression profiles, which are further changed by the diffusion of signaling 64 factors, thus generating a feedback loop of cell and tissue differentiation (Turing, 1952; Marcucio 65 et al., 2005; Meinhardt, 2008; Franz-Odendaal, 2011; Xu et al., 2015). 66 These signaling factors may target specific cell lineages, but the contact between neighboring 67 tissues may produce correlated changes between them due to mechanical interactions and through mutual signaling cascades that induce specific behaviors, and such interactions are necessary for the proper development of both tissues (Cheverud et al., 1992; Ravosa et al., 2000; Jiang et al., 2002; Marcucio et al., 2011). For instance, cranial Vault growth, which occurs on the final stages of pre-natal development, is a result of the tension exerted by the growing brain on

its encircling membranes, inducing them to secrete signaling factors which promote bone growth (Opperman, 2000; Rice et al., 2003). This mechanism promotes a tight association between the Vault and the brain. Furthermore, post-natal facial growth is induced by muscular activity related to masticatory function, and this effect dominates post-natal skull growth (Zelditch et al., 1992; Herring, 2011). Muscle activity mainly affects the focal Oral region in which muscular 77 forces are exerted, but other skull regions to which these muscles are attached are also affected to a lesser extent, such as the Zygomatic and Vault. Therefore, development is composed of 79 a series of such events, and it may be difficult to isolate the effect each individual process has 80 on covariance structure, given their spatial overlap (Hallgrimsson et al., 2009); hence, both the 81 temporal hierarchy and the spatial organization of developmental and functional interactions influence the patterns embedded in genetic covariance structure. 83 Empirical evidence on long-term changes on genetic covariance structure (comparative quantitative genetics; Steppan et al., 2002) rely on the correspondence between G and its phenotypic counterpart P, because estimating G demands high sample sizes and nonetheless such estimate is prone to substantial error, since sample units are families rather than individuals (Meyer, 1991; Houle & Meyer, 2015). The similarity in covariance structure between P and G ("Cheverud's Conjecture"; Cheverud, 1988; Roff, 1995), has been supported in different trait systems and organisms (e.g.: Waitt & Levin, 1998; Dochtermann, 2011; Garcia et al., 2014), indicating that 90 environmental effects (E) are either uncorrelated or exhibit patterns similar to G, since these 91 effects exert their influence upon phenotypes through the same developmental processes by 92 which genetic variation is structured (Rice, 2002, 2004). The comparative analysis of phenotypic covariance structure shows that P can remain stable 94 in macroevolutionary scales (e.g. Marroig & Cheverud, 2001; Oliveira et al., 2009b; Kolbe et al., 2011). This stability may be a consequence of the alignment between phenotypic covariance structure and the local features of the adaptive landscape acting over different lineages. In New World Monkeys (Platyrrhini), divergence in body size among lineages is closely associated with shifts in diet composition (Rosenberger, 1992). Size represents the main feature of both genetic

(Cheverud, 1995, 1996b) and phenotypic (Marroig & Cheverud, 2005) covariance structure, and for some groups within New World Monkeys, such as Atelids and Callithrichines, divergence in 101 body size is a direct consequence of directional selection (Marroig & Cheverud, 2010; Marroig 102 et al., 2012). On the other hand, dietary or locomotory divergence between sister lineages can 103 rearrange the patterns expressed in P (Young & Hallgrimsson, 2005; Young et al., 2010; Monteiro 104 & Nogueira, 2010; Haber, 2015), and such reorganization hints at changes in the underlying 105 architecture of the genotype/phenotype map, indicating that the patterns privileged by selection 106 can overcome constraints imposed by development (Jamniczky & Hallgrímsson, 2009). 107 The adaptive radiation of phylostomid bats, for instance, has involved rearrangements of 108 mandibular phenotypic covariance structure (Monteiro & Nogueira, 2010), as the dietary di-109 vergence in this group from a probable insectivore ancestor towards more specialized diets 110 (such as frugivory or sanguivory) imply distinct functional relationships among mandibular components. This reorganization may indicate heterogeneity in the non-linear aspects of the 112 adaptive landscape for different phylostomid lineages. For bats in general, their specialized 113 locomotory behavior is associated with the decoupling between fore- and hindlimb covariance structure when compared to other mammalian lineages (Young & Hallgrímsson, 2005). To a 115 lesser extent, a similar pattern can be observed in hominoids when compared to remaining 116 Anthropoids (Young et al., 2010). Another example of divergence in covariance structure among 117 lineages due to changes in functional demands are those observed in ruminants (Haber, 2015), 118 which may related to increased metabolic requirements for occupying open habitats, as the 119 divergence between Bovidae and Cervidae is associated with increased Nasal integration in the 120 former. Thus, divergence among sister lineages may imply a reorganization of both genetic and 121 phenotypic covariance structure of phenotypic traits if such divergence is associated with distinct 122 functional relationships, represented by the non-linear components of adaptive landscape. 123 Currently, there are several methods dedicated to the comparison of covariance structures (e.g. 124 Krzanowsky, 1979; Phillips & Arnold, 1999; Cheverud & Marroig, 2007), and such methods are often focused on constructing hypothesis of similarity or dissimilarity; However, it is not always

clear which hypothesis is the adequate one, and frequently either hypotheses can be rejected for the same pair of matrices, hindering interpretation (Ackermann & Cheverud, 2000; Marroig & 128 Cheverud, 2001; Haber, 2015). By expressing covariance matrix (dis)similarities with a single 129 metric, these methods lack an explicit way of describing structural differences between pairs 130 of covariance matrices. Furthermore, these methods also lack a direct manner to incorporate 131 phylogenetic relatedness in pairwise comparisons. While extensions for these methods have 132 been proposed to deal with the first limitation (e.g. Hansen & Houle, 2008; Hine et al., 2009; 133 Marroig et al., 2011), the second issue is usually resolved by comparing the set of pairwise matrix 134 comparisons with the set of phylogenetic distances among lineages (e.g. Marroig & Cheverud, 135 2001; Oliveira et al., 2009b). Although such comparison provides a first approximation to 136 this problem, it has the same problems as comparing covariance matrices themselves, that is, 137 summarising the relationship between patterns expressed in P or G in a set of lineages and 138 their phylogenetic relatedness to a single value. In this manuscript, we explore novel methods to 139 circumvent these issues.

1 Objectives

The interplay between developmental and functional interactions, and their relationships with the 142 topology of adaptive landscapes may influence the divergence between covariance structure in 143 sister lineages. In the present work, we evaluate the stability of phenotypic covariance structure in 144 skull size and shape along the diversification of Anthropoid Primates. We build upon approaches 145 proposed by other authors (Marroig et al., 2011; Aguirre et al., 2013; Haber, 2015) in order to 146 explicitly incorporate phylogenetic relationships into the comparative analysis of covariance 147 structure, under the hypothesis that different cranial regions will exhibit different degrees of 148 stability among sister lineages, thus producing a non-random pattern of changes in covariance 149 structure.

Methods

Sample

Our sample consists of 5108 individuals in 109 species, distributed throughout all major Anthropoid clades above the genus level, comprising all Platyrrhini genera and a substantial portion 154 of Catarrhini genera. We associate this database with a ultrametric phylogenetic hypothesis for 155 Anthropoidea (Figure S2), derived from Springer et al. (2012). 156 Individuals in our sample are represented by 36 landmarks; these landmarks were registered 157 using a Polhemus 3Draw and a Microscribe 3DX for Platyrrhini and Catarrhini, respectively. 158 Twenty-two unique landmarks represent each individual (Figure S1), since fourteen of the 36 159 registered landmarks are bilaterally symmetrical. For more details on landmark registration, see 160 Marroig & Cheverud (2001) and Oliveira et al. (2009b). Databases from both previous studies 161 were merged into a single database, retaining only those individuals in which all landmarks from 162 both sides were present. In the present work, we considered only covariance structure for the 163 symmetrical component of variation; therefore, prior to any analysis, we controlled the effects 164 of variation in assymmetry. We followed the procedure outlined in Klingenberg et al. (2002) for bilateral structures by obtaining for each individual a symmetrical landmark configuration, 166 averaging each actual shape with its reflection along the sagittal plane. 167 We used this database to obtain local shape variables (Márquez et al., 2012), which represent in-168 finitesimal volumetric expansions or retractions, calculated as the natural logarithm determinants 169 of derivatives of the thin-plate spline between each individual in our sample and a reference 170 shape (in our case, the mean shape for the entire sample, estimated from a Generalized Pro-171 crustes algorithm). Such derivatives were evaluated at the midpoints between pairs of landmarks 172 represented in Figure S1, for a total of 38 local shape variables. 173 After obtaining these values, we estimated covariance P-matrices for size (represented by the

natural logarithm of Centroid Size) and local shape variables after removing fixed effects of little

interest here, such as sexual dimorphism, for example. These effects were removed through a multivariate linear model adjusted for each species, according to Figure S2. We adjusted 177 such models under a Bayesian framework, sampling 100 residual covariance matrices from 178 the posterior distribution of each model. These distributions allow us to estimate uncertainty 179 for any parameters derived from these matrices as credibility intervals; furthermore, since 180 posterior distributions are conditional upon the prior distribution we used — a uniform Wishart 181 distribution — every matrix sampled from these posterior distributions is also a realization 182 of a Wishart distribution, therefore positive-definite regardless of sample size (Gelman et al., 183 2004). In this framework, lower sample sizes imply in broader and less informative credibility 184 intervals. For each posterior sample, we estimated geometric mean covariance matrices, since this 185 mean respects the underlying geometry of the Riemannian manifold in which positive-definite 186 symmetric matrices lie (Moakher, 2005, 2006). These mean P-matrices are also positive-187 definite, regardless of sample size. For each species, we ran independent models, with 13000 iterations of MCMC sampling, discarding the 3000 initial iterations as a burn-in period and further sampling one covariance matrix per 100 iterations to avoid autocorrelations induced by 190 sequential sampling.

Phylogenetic Decompostion of Matrix Diversity

In order to evaluate the distribution of covariance structure diversity during Anthropoid diversification, we estimated Riemannian distances among all pairs of mean P-matrices, according to the definition given by Mitteroecker & Bookstein (2009); for any pair of positive-definite covariance matrices C_i and C_j of size $p \times p$, the distance $d(C_i, C_j)$ is given by

$$d(\mathbf{C}_i, \mathbf{C}_j) = \sqrt{\sum_{k=1}^p \ln^2 \lambda_k(\mathbf{C}_i \mathbf{C}_j^{-1})}$$
 (1)

where $\lambda_k(\cdot)$ refers to the k-th eigenvalue obtained from the spectral decomposition of a given matrix, in this case the product $\mathbf{C}_i\mathbf{C}_j^{-1}$. This distance among pairs of P-matrices is negatively

correlated with Random Skewers comparisons (Figure S3), a measurement of matrix similarity

explored elsewhere (Cheverud & Marroig, 2007). The similarity between Riemannian distances and Random Skewers similarity indicate that our conclusions would be the same regardless of the metric used to characterize matrix similarity or dissimilarity.

Using these distances among P-matrices, we estimate matrix diversity at each node of the phylogenetic tree of Anthropoidea using a measurement of the weighted distance among the distributions of matrix distances for the two descending edges, based on Pavoine $et\ al.\ (2010)$.

For a fully resolved tree, diversity w_i on node i is given by

$$w_i = \frac{1}{2} \frac{n_\alpha n_\beta}{n_i n_T} D_\Delta^2(P_\alpha, P_\beta) \tag{2}$$

where α and β represent the subsets of descendants from node i, and n refers to the number of species on each set (n_i for the total descendants of node i; n_T for the total number of species considered; n_{α} and n_{β} for the size of descending subsets). $D_{\Delta}(P_{\alpha}, P_{\beta})$ represents the actual distance between the two distributions P_{α} and P_{β} for descending nodes, as formulated by Rao (1982):

$$D_{\Delta}(P_{\alpha}, P_{\beta}) = \sqrt{2\left(2H_{\Delta}\left(\frac{P_{\alpha} + P_{\beta}}{2}\right) - H_{\Delta}(P_{\alpha}) - H_{\Delta}(P_{\beta})\right)} \tag{3}$$

212 where

$$H_{\Delta}(P) = \sum_{i,j \in P} \frac{d^2(\mathbf{C}_i, \mathbf{C}_j)}{2} \tag{4}$$

represents Rao's quadratic entropy among Riemannian distances $d(\mathbf{C}_i, \mathbf{C}_j)$ as defined in Equation 1.

Following the framework estabilished by Pavoine *et al.* (2010), diversity w_i can be normalized as $v_i = w_i / \sum_i w_i$ to represent the percentage of diversity with respect to the total diversity on

the phylogenetic tree. We test three different hypothesis regarding the distribution of v_i values through Anthropoid diversification: (1) that P-matrix diversity is concentrated in a single node; (2) that P-matrix diversity is concentrated in a reduced number of nodes; (3) that P-matrix diversity is skewed towards either the root or tips of the phylogeny, in a two-tailed test. We test 220 each hypothesis against the null hypothesis that the distribution of matrix diversity is randomly 221 arranged over the phylogeny; such null hypothesis is represented by randomizing the association 222 between terminal branches and covariance matrices, constructing 9999 distributions of v_i values 223 that represent this scenario. Each test is carried out using a different parameter derived from the 224 distribution of v_i values (described in detail by Pavoine et al., 2010), comparing the actual value 225 obtained with a null distribution constructed using permutations. The third hypothesis can be 226 tested either by considering only the topology of the tree and by also considering branch lengths; 227 both tests are similar to Blomberg's (2003) K test, as they search for a phylogenetic signal in 228 covariance structure diversity. 229

230 Characterizing Covariance Matrix Variation

The tests described in the previous section allow us to pinpoint which nodes contribute mostly 231 to divergence in covariance structure; however, these tests are not designed to properly describe 232 the actual changes in P-matrix structure that are responsible for such divergence. To actually 233 represent these changes in a comprehensible manner, we combine a number of ordination 234 techniques to reduce the dimensionality of the manifold that contains covariance matrices of size 235 $p \times p$ (Figure 1). 236 For a Riemannian manifold, there exists at least one bijective function defined in the neighbourhood of a given covariance matrix M that maps the manifold to an Euclidean space — a 238 hyperplane with p(p-1)/2 dimensions also contained in $\mathbb{R}^{p\times p}$ — and equips the manifold with a 239 notion of inner product, thus allowing the construction of an orthonormal basis that can be used 240 to describe variation in P-matrix structure. For a covariance matrix X in the neighbourhood of 241

242 M,

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$$f(\mathbf{X}) = \log(\mathbf{M}^{-\frac{1}{2}}\mathbf{X}\mathbf{M}^{-\frac{1}{2}}) \tag{5}$$

represents one possible function. Here, the logarithm operator refers to matrix logarithm; for symmetric positive-definite matrices, this transformation is equivalent to applying the usual logarithm function to the eigenvalues of such matrix and reverting the spectral decomposition. The function defined in Equation 5 also transforms the Riemannian distance among covariance matrices defined in Equation 1 into Euclidean distances between transformed matrices (Moakher, 2005). We defined the average matrix among all sampled P-matrices as the location parameter M 249 to map the entire set of posterior P-matrices into an Euclidean space. We then used these 250 P-matrices to produce axes of matrix variation using an eigentensor decomposition (Basser & 251 Pajevic, 2007; Hine et al., 2009) obtaining a set of eigentensors and eigenvalues that summarise 252 matrix variation (Figure 1a). As a consequence of using the mean covariance matrix for the entire 253 sample over Equation 5, the projections over eigentensors we obtained are naturally centered on 254 M. 255 We used the projections of P-matrices over these eigentensors as traits in a phylogenetic 256 Principal Component Analysis (pPCA; Jombart et al., 2010), which produces a new set of axes of matrix variation that considers both trait dispersal and phylogenetic relationships among species 258 simultaneously. If Z represents a matrix with projections of n P-matrices over each eigentensor 259

$$\frac{1}{2n}\mathbf{Z}^t(\mathbf{W} + \mathbf{W}^t)\mathbf{Z} \tag{6}$$

where W represents the matrix of phylogenetic distances between species; here, the distance w_{ij} between tips i and j is the sum of branch lengths from their last common ancestor to both

on its columns, phylogenetic PCs are the eigenvectors obtained from

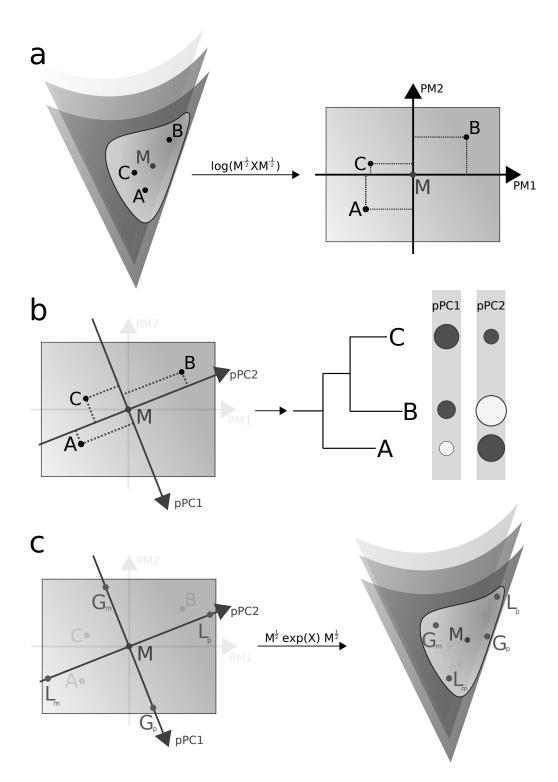


Figure 1: Representation of the steps used to characterize covariance matrix variation. In (a), the set of covariance matrices A, B and C in the neighbourhood of M are projected into an Euclidean space and eigentensors are estimated (PM1 and PM2); in (b), these eigentensors are rotated to incorporate phylogenetic relatedness; in (c), covariance matrices at the upper and lower bounds of the confidence intervals for each axis are returned back to the original manifold. See text for more details.

tips. Other definition of phylogenetic distances may be used (see Jombart et al., 2010); the results

we show here were not changed by considering different measures of phylogenetic distance, 264 regardless of whether these distances consider branch lengths among species or not. 265 Such analysis produces both positive and negative eigenvalues, which are respectively associated 266 with variation close to the root of the tree ('Global') and variation close to the tips ('Local') in 267 matrix structure (pPC1 and pPC2 in Figure 1b, respectively). Pavoine et al. (2010) argues that 268 this contrast between Global and Local components in phylogenetic PCs reflects phylogenetic 269 signal and convergence in trait values, respectively, as observed in the distribution of Moran's 270 (1948, 1950) autocorrelation Indexes for each axis constructed in this manner. This index can 271 be understood as the degree onto which an observed value in a given species is determined 272 by the values on its phylogenetic "neighborhood" (as expressed by W), in a similar manner to 273 autoregressive models (Cheverud & Dow, 1985; Cheverud et al., 1985). 274 For each pPC obtained in this manner, we obtained two covariance matrices by estimating the 275 upper and lower limits of the 95% confidence interval for each axis and mapping these values 276 back to the manifold of symmetric positive-definite matrices (Figure 1c), defining the inverse

operation associated with Equation 5 as

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$$f^{-1}(\mathbf{X}) = \mathbf{M}^{\frac{1}{2}} \exp(\mathbf{X}) \mathbf{M}^{\frac{1}{2}}$$
 (7)

to describe matrix variation associated with each axis comparing each pair of matrices with the
Selection Response Decomposition tool (Marroig et al., 2011), in order to pinpoint which traits
are associated with the divergence in covariance structure associated with each pPC.

In order to characterize such divergence in covariance structure with respect to the uncertainty
in P-matrix estimation, we carried out the analyses described in this section with both mean
P-matrices obtained from posterior samples, and with posterior samples themselves, obtaining
the posterior samples are accounted to the analyses described in this section with both mean
P-matrices obtained from posterior samples, and with posterior samples themselves, obtaining

where the exponential operator refers to matrix exponential. We used these covariance matrices

allowing us to estimate a posterior distribution of mean SRD scores for each trait in all pPCs. We

use the phylogenetic PCA estimated over mean P-matrices in order to represent the phylogenetic

289 patterns described by each pPC.

290 We use the posterior distribution of mean SRD scores over traits and pPC to investigate whether

291 these changes in trait-specific covariance structure along Anthropoid diversification are randomly

292 distributed with respect to the skull regions delimited in Table S2 by comparing SRD scores

estimated within each region for all pPCs. The association between cranial traits and such

regions reflect their functional significance and developmental origins.

5 Software

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²⁹⁶ We performed all analyses under R 3.2.1 (R Core Team, 2015). We fitted Bayesian linear models

for estimating posterior P-matrix samples using the MCMCglmm package (Hadfield, 2010). Both

eigentensor decomposition and the SRD method are provided by the evolgg package (Melo et

299 al., 2015); the phylogenetic decomposition of diversity was provided by Pavoine et al. (2010) in

their Supplemental Material, while pPCA is implemented in the adephylo package (Jombart &

Dray, 2010). In order to obtain symmetrical landmarks configurations, we used code provided by

Annat Haber, available at http://life.bio.sunysb.edu/morph/soft-R.html.

Results

The distribution of covariance matrix diversity along the Anthropoid phylogeny (Figure 2a)

305 indicates that the divergence between Catarrhini and Platyrrhini contributes to approximately

306 10% of all covariance matrix diversity; within New World Monkeys, the divergence between

Atelidae and Cebidae contributes with 4% to covariance structure diversity, while for Catarrhines,

the divergence between Hominoidea and Cercopithecoidea contributes 3% to overall covariance

matrix diversity. The remaining P-matrix diversity is distributed along the tree, with a consistent

decay of explained diversity the closer any given node is from terminal branches. With respect to
the hierarchy of tests exploring the phylogenetic distribution of matrix diversity (Table 1), all tests
reject their null hypotheses of random arrangements of diversity along the Anthropoid phylogeny,
thus indicating that covariance matrix diversity exhibits some degree of phylogenetic structure,
with these few basal nodes contributing to a greater extent to such diversity.

Table 1: Phylogenetic decomposition of covariance matrix structure diversity.

	Value	$\mathbf{Expected}^a$	$\mathbf{Distance}^b$	<i>p</i> -value
Single Node	0.106	0.029	13.455	< 10 ^{−4}
Few Nodes	0.248	0.139	13.545	$< 10^{-4}$
Root/Tip Skewness ^c	0.632	0.505	12.197	$< 10^{-4}$
Root/Tip Skewness ^d	0.381	0.505	-11.067	$< 10^{-4}$

^a refers to the distribuition of permutated values;

The eigenvalue distribution of phylogenetic Principal Components that describe P-matrix structure (Figure 3) shows that individual Global components surpass their Local counterparts in terms of explained variance, such that the first Global component has a larger contribution 317 than the first Local component to interspecific P-matrix variation. While there are less positive 318 than negative eigenvalues, Global components explain more than half (57%) of the total P-matrix 319 variation, considering mean absolute eigenvalues obtained from their posterior distribution. The 320 posterior distribution of Moran's Index for phylogenetic Principal Components (Figure 4) is 321 assymetric towards positive values, also indicating that the similarity produced by phylogenetic 322 inertia is greater than the similarity produced by convergence in P-matrix structure. 323

Considering the distribution of matrix projections for each species on these pPCs (Figure 2b),
we observe that the first Global pPC separates New World and Old Monkeys, while the second
Global pPC consists of a contrast between Atelids and Cebids, and the third Global pPC
generally contrasts Hominids with the remaining Anthropoids, thus indicating a pattern of
P-matrix variation consistent with those observed in the diversity decomposition summarized

^b difference between empirical and expected values in standard deviations of the permutated values distribution;

^c considering only topology;

^d including branch lengths.

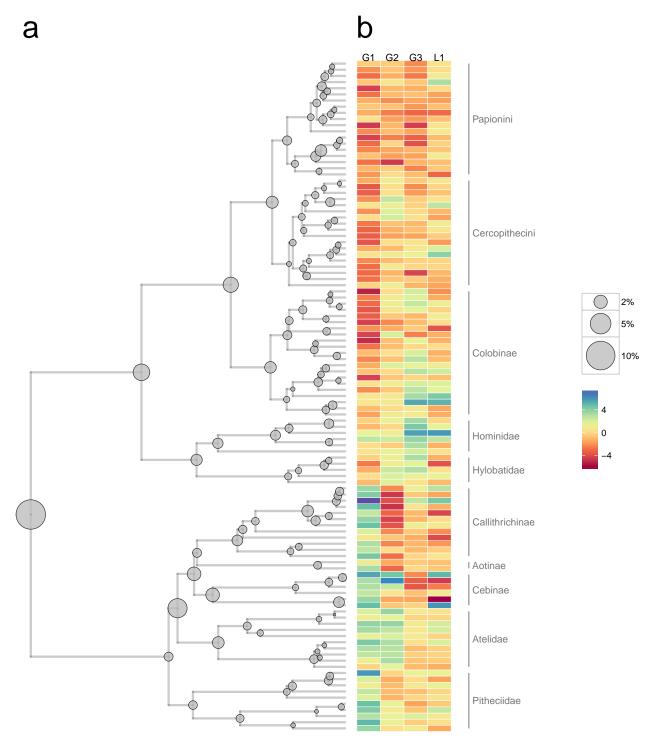


Figure 2: Phylogenetic decomposition of P-matrix variation. (a) Decomposition of matrix diversity over the phylogenetic hypothesis for Anthropoidea; the size of each circle indicates the percentage of diversity on each node, according to the legend. (b) Mean P-matrices of each species projected over the first three and the last phylogenetic Principal Components (G1-3 and L1, respectively); cell colors represent projection values, according to the legend.

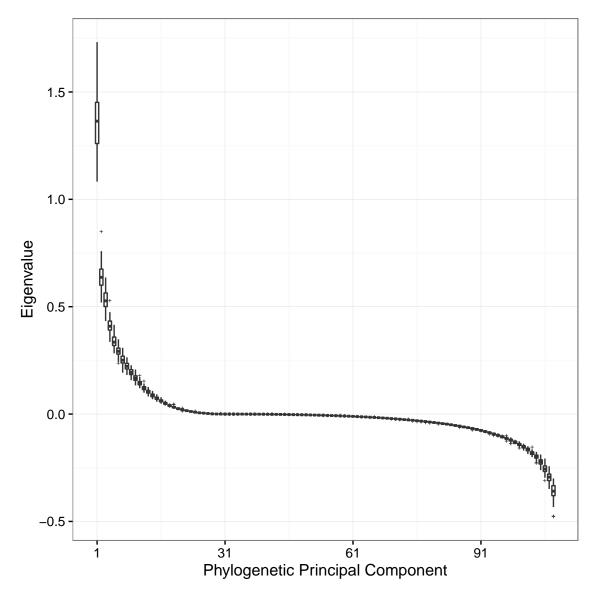


Figure 3: Posterior distribution of eigenvalues obtained for pPCs. Positive eigenvalues are associated with phylogenetic signal; negative eigenvalues are associated with convergence.

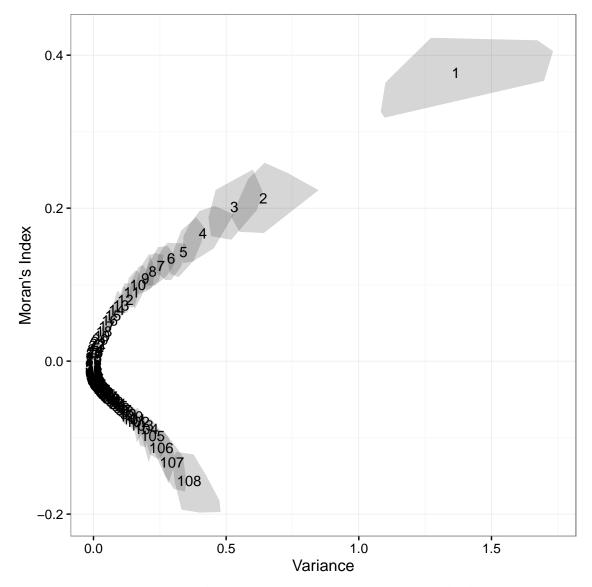


Figure 4: Posterior bivariate distribution of variances explained by each phylogenetic PC *versus* Moran's Indexes estimated for each axis.

in Figure 2a. The first Local pPC consists of localized contrasts between sister species, such as the two representatives of *Saimiri*, for instance; however, matrix projections over this axis can be explained by the effect of sample sizes (Figure S4). These localized contrasts can thus be explained on the account of substantial sample size differences between sister species.

The variation in trait-specific covariance structure described by these four phylogenetic Principal 333 Components, as captured by comparing the posterior distribution of confidence intervals for 334 each axis using Selection Response Decomposition (Figure 5) indicates that, regardless of which 335 axis is considered, traits with lower posterior SRD scores are usually localized in either Orbit or 336 Basicranium, along with log Centroid Size, which represents the covariance structure associated 337 with allometric relationships. Traits in remaining skull regions (Oral, Nasal, Zygomatic, Vault) 338 consistently exhibit higher SRD scores, thus indicating a more stable covariance structure 339 associated with these regions throughout Anthropoid diversification. 340

The overall distribution of average posterior SRD scores along the entire set of phylogenetic PCs (Figure S5) indicates that this behavior detailed on Figure 5 for G1-3 and L1 is the norm also for the remaining pPCs, that is, Orbit and Basicranial traits along with allometric relationships consistenly have lower average SRD scores than other skull regions. Notice that the explained amount of variance for intermediate pPCs (roughly from pPC 31 to 61) is very close to zero (Figure 3); thus, although matrices describing confidence intervals in these axes can be recovered, the pattern expressed by them should not be taken into account.

Discussion

Since its conception, the hypothesis that functional interactions among morphological traits
shape their phenotypic covariance structure (Olson & Miller, 1958) has been complemented
with the notion that developmental interactions mediate the functional relationships among
traits dynamically. This means that it is difficult to separate the relative contribution of either
development or function to phenotypic integration (Cheverud, 1996a), especially considering

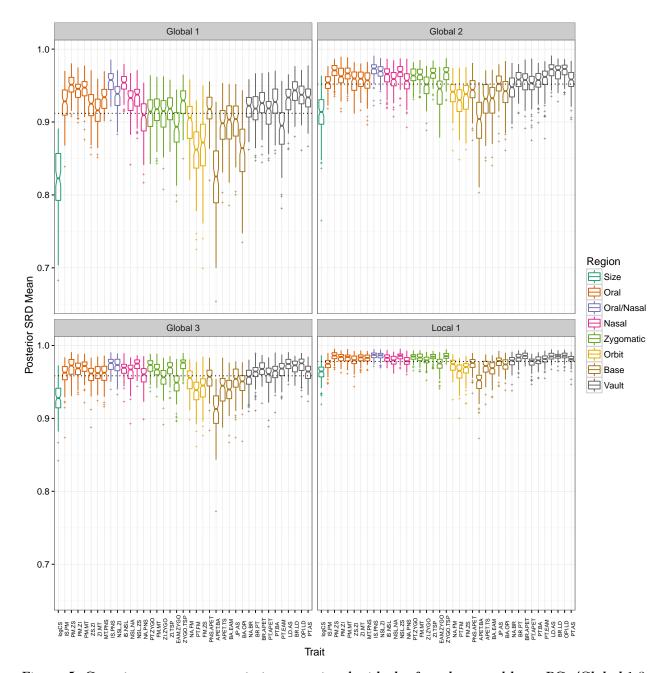


Figure 5: Covariance structure variation associated with the first three and last pPCs (Global 1-3 and Local 1, respectively), represented using posterior mean SRD scores. Dotted lines represent average SRD scores for each comparison. Traits are colored according to their association with each skull region, according to the legend.

that the structure of developmental interactions is thought to match the pattern of optimal functional interactions (Lande, 1980; Cheverud, 1984; Wagner, 1996; Jones, 2007), which further entangles both phenomena. Thus, changes in covariance structure between sister lineages should be associated with the interplay between functional and developmental interactions.

Our results indicate that the changes in phenotypic covariance structure associated with all 358 phylogenetic principal components follow a similar pattern with respect to changes in trait-specific 359 covariances (Figures 5 and S5). These changes are mostly associated with either Basicranial and 360 Orbital trait sets, along with allometric relationships, represented by the covariances between log 361 Centroid Size and local shape variables. While the comparison between covariance matrices in 362 the limits of each phylogenetic PC represents the overall pattern of dissimilarity in trait-specific 363 covariance structure associated with each axis, the actual values obtained represent only a 364 fraction of the overall divergence between Anthropoid lineages. Given any two P-matrices 365 associated with a pair of species, the actual SRD comparison between them will be some linear 366 combination of their divergences along each pPC axis. We now attempt to interpret these results 367 in terms of the recent advances in developmental biology for each cranial region.

Orbital traits are mostly associated with the development of the postorbital bar in Euprimates, 369 as opposed to Plesiadapiforms, and in Anthropoids it fully develops into the orbital cavity and 370 postorbital wall (Ravosa & Savakova, 2004). The origins of this structure have been linked to 371 the distribution of masticatory loadings around the comparatively large primate eye for the 372 maintenance of a stable, forward-facing visual field even during feeding behavior (Ravosa et 373 al., 2000). Although most of the cranial Vault originates from intramembranous ossification 374 induced by the growing brain and thus derived from the neural crest, the influence of mesoderm-375 derived condensations and its pattern of endochondral ossification is necessary for the proper 376 development of the fronto-nasal and fronto-zygomatic sutures, which affect the brow ridge and 377 both medial and lateral orbital walls (Jiang et al., 2002). 378

In the same manner, the Basicranium originates from a set of thirteen condensations derived from

both precursors, which exhibit a mosaic pattern of both endochondral and intramembranous ossification (Lieberman, 2011). Such processes occurs early during skull development, and 381 the spatial overlap of developmental processes (Hallgrímsson et al., 2009) may also explain 382 Basicranium variation in covariance structure. Moreover, the angulation between anterior and 383 posterior elements of the Basicranium has significantly changed during primate evolution, and 384 such property appears to have evolved in coordination with Facial growth relative to the cranial 385 Vault, accomodating both structures on each other (Scott, 1958; Lieberman *et al.*, 2000, 2008). 386 Therefore, both Orbit and Basicranium are located in the boundary between domains of two 387 precursor tissues that originate osteological elements in the skull, and their proper development 388 is affected by both of them; furthermore, their functional significance is also associated with the 389 accommodation of remaining structures. These characteristic properties of their development may 390 thus be sufficient to explain their divergence among lineages in terms of covariance structure. 391 On the other hand, the structure of phenotypic covariances for Oral, Nasal and Vault traits remained stable through the course of Anthropoid diversification. These regions exhibit more 393 consistent patterns of developmental processes, when compared to the Orbit or Basicranium. Both Oral and Nasal regions exhibit a great degree of interactions with soft tissue during prenatal development; the Oral region further suffers the influence of muscleskeletal interactions 396 associated with Facial growth in postnatal development, which also contributes to its pattern of 397 stability in covariance structure (Zelditch et al., 1992; Herring, 2011; Lieberman, 2011). The cranial 398 Vault exhibits a more regular pattern of growth, induced by the underlying brain (Lieberman, 399 2011; Esteve-Altava & Rasskin-Gutman, 2014). While there are prenatal developmental processes 400 associated with the integration of both structures (Marcucio et al., 2005, 2011) and postnatal 401 muscle-bone interactions may be understood as a overall integrating factor — since the skull as 402 whole is affected by such interactions — each of these regions is located within the bounds of 403 precursor components, thus exhibiting a stable association between developmental processes 404 that originate these regions and their functional aspects. 405

6 Finally, allometric relationship exhibit changes in covariance structure across phylogenetic

principal components with magnitudes equivalent to Basicranial traits (Figures 5 and S5). While such relationships are explored elsewhere (Garcia et al., in prep), it is noteworthy that these 408 changes may be associated with the role of directional selection for body size, which shaped 409 the extant diversity of New World (Marroig & Cheverud, 2005, 2010) and Old World Monkeys 410 (Cardini et al., 2007; Cardini & Elton, 2008), as well as the more complex relationships between 411 size and shape observed within Hominidae (Ackermann & Cheverud, 2004; Mitteroecker et al., 412 2004; Mitteroecker & Bookstein, 2008; Schroeder et al., 2014). To some extent, selection for body 413 size produces mean shape differences among lineages in a manner consistent with the ancestral 414 allometric relationships (Lande, 1979; Schluter, 1996), but selective pressures for size may alter 415 such relationships depending on the structure of interactions between size, shape and ontogeny 416 (Pélabon et al., 2013, 2014). 417 The phylogenetic distribution of changes in covariance strucuture (Figure 2) reveals an even 418 distribution of such changes throughout Anthropoid diversification, with three different instances in which more substantial changes have occurred, in order of the estimated P-matrix diversity at 420 each point (Figure 2a): the divergence between New World and Old World Monkeys, between Atelidae and Cebidae, and between Hominoidea and the remaining Anthropoids. Results from the phylogenetic principal component analysis (Figure 2b) recover a similar pattern, considering 423 the contribution of these lineages to each axis of matrix variation, and the posterior distributions 424 of the associated eigenvalues for the first, second and third phylogenetic principal components 425 (Figure 3) indicate that these cladogenetic events can be set apart in terms of their relative 426 contributions to P-matrix diversity. However, the apparent greater contribution of the divergence 427 between Atelids and Cebids to P-matrix diversity may be a consequence of the phylogenetic 428 structure imposed by either analyses, given that the actual pattern of divergence in the third 429 phylogenetic component (Figure 2b) puts the Hominoid lineage apart not only from its sister 430 group, Cercopithecoidea, but also from remaining lineages. Since the third pPC is not a proper 431 constrast between sister lineages, but rather between Hominoidea and the paraphyletic grouping 432 obtained from removing this group, the imposed phylogenetic structure of both models might 433

reduce the contribution of this lineage to overall P-matrix diversity in either analyses.

Nonetheless, both events are overshadowed by the separation between New World and Old 435 World Monkeys, since this divergence is associated with an unequivocally higher eigenvalue, as 436 indicated by their posterior distribution (Figure 3). Considering that the migration from Africa 437 to South America probably occured through the colonization of successive island environments 438 that may have occurred in the South Atlantic Ocean during the Eocene (Oliveira et al., 2009a), 430 the effective population sizes in the ancestral population for all Platyrrhines probably plummeted 440 during this vicariant event. Thus, the subtle differences in P-matrix diversity such separation 441 represents may indicate that changes in the genetic architecture of cranial traits due to drift 442 and/or founder effects (Goodnight, 2000; Brito et al., 2005) may be a null hypothesis with 443 enough explanatory power for this divergence, against an alternative hypothesis of differences in covariance structure between Platyrrhines and Catarrhines due to either directional or stabilizing 445 selective pressures. Although these hypotheses could be tested using comparative methods, for instance (e.g. Haber, 2015), it is unclear at this point whether these methods are adequate to test hypothesis with respect to the evolution of morphological integration, given that they were conceived to model the evolution of mean phenotypes under the general assumption of constant 449 genetic variances (Hansen & Martins, 1996). The descriptive approach we use here, focussing 450 on comparing credible intervals between covariance matrices while incorporating phylogenetic 451 structure has the advantage of imposing minimal assumptions. However, regardless of whether 452 comparative methods are suited to deal with the evolution of morphological integration or not, 453 addressing this question requires adequate sampling both within and between lineages (Melo et 454 al., 2015). 455 The pattern recovered by the first Local component (lower right panel of Figure 5) also describes 456 a pattern similar to that of Global components. However, the distribution of lineages in this 457 component scales with the logarithm of sample sizes (Figure S4), thus indicating the influence 458 of sampling effects upon this component. Basicranial and Orbital traits exhibit overall lower 459 covariances, both among themselves and between other regions, probably owing to the influence

of multiple and disparate developmental processes; however, such lower correlation values also imply more uncertainty in their estimation. On the other hand, it is also important to consider 462 that these directions depicted as changes in trait-specific covariance structure using the SRD 463 method are orthogonal in the Euclidean image of the exponential-logarithm map. Thus, these 464 components represent different aspects of covariance divergence among lineages, which in this 465 particular case are concentrated in the same traits regardless of the phylogenetic scale considered. 466 The association between the first Local component with sample sizes hinders its interpretation as 467 biologically meaningful, given that Local components should be related to convergence among 468 phylogenetically disparate lineages (Jombart et al., 2010); rather, this component seems to be 469 associated with divergence in the statistical properties of estimated covariance matrices for sister 470 lineages. 471 Considering both the pattern associated with the first Local component and the prevalence 472 of variation associated with Global components over Local components (Figures 3 and 4), the 473 contribution of historical constraints is greater than the contribution of convergence among 474 disparate lineages. Given that both sets of components are associated with the same structure 475 of changes in covariance structure (Figures 5 and S5) related to some of the properties of 476 skull development, these results highlight the overall stability of shape covariance structure in 477 Anthropoids, in agreement with previous works using traditional morphometrics (Marroig &

480 Conclusions

Cheverud, 2001; Oliveira *et al.*, 2009b).

The stability of shape covariance structure in Anthropoids may be a consequence of either constraints on mammalian skull development or the prevalence of a constant pattern of functional relationships imposed by stabilizing selection; such dichotomy has already been pointed out by other authors (e.g. Marroig & Cheverud, 2001; Porto *et al.*, 2009; Oliveira *et al.*, 2009b). Here, we favor the second point of view, in light of the evidence for the evolution of genetic or phenotypic covariances under directional selection (Jones *et al.*, 2012; Melo & Marroig, 2015;

Careau et al., 2015) or under relaxation of stabilizing selection (Jamniczky & Hallgrímsson, 2009). Furthermore, the available comparative data (Monteiro & Nogueira, 2010; Haber, 2015) 488 indicate that some mammal lineages, such as phylostomid bats and ruminants, have diverged in 489 phenotypic covariance structure probably due to changes in the adaptive landscapes resulting 490 from differences in ecological processes acting on different lineages. On the other hand, for 491 Anthropoids, their stability in P-matrix structure may thus be possible as a consequence of 492 stability of the functional relationships imposed over by the adaptive landscape in this lineage, 493 as it seems that in this group, dietary shifts produced changes on body size only, at least for 494 Platyrrhines (Marroig & Cheverud, 2010). 495 From an information theory point of view (Brooks et al., 1989; Frank, 2009), natural selection 496 will increase the correlation between information encoded in a population and the information 497 that represents its environment, and the suggestion that developmental systems share properties 498 with machine learning algorithms (Watson et al., 2014) only reinforces such view. Furthermore, 499 the regulation of developmental systems, through both genetic and epigenetic effects, may also 500 be targeted by selection for both robustness and replicability (Hansen, 2011). These different ways of thinking about developmental and functional interactions mean that the probabilistic 502 distribution of phenotypes, as a consequence of either genetic information or epigenetic effects, 503 will tend to match the properties of the distribution of fitness in a particular environment. Thus, 504 the maintenance of the same phenotypic distribution over macroevolutionary timescales indicates 505 that the fitness distribution itself may be stable as well. 506

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