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# Q1 Is complex allometry in field metabolic rates of mammals a 2 statistical artifact?

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#### ABSTRACT

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lometry in which the exponent in a simple, two-parameter power equation increases steadily as a depen-18 dent function of body mass. The reports were based, however, on indirect analyses performed on 19 logarithmic transformations of the original data. I re-examined values for FMR and body mass for 114 spe-20 cies of mammal by the conventional approach to allometric analysis (to illustrate why the approach is un-21 reliable) and by linear and nonlinear regression on untransformed variables (to illustrate the power and 22 versatility of newer analytical methods). The best of the regression models fitted directly to untransformed 23 observations is a three-parameter power equation with multiplicative, lognormal, heteroscedastic error 24 and an allometric exponent of 0.82. The mean function is a good fit to data in graphical display. The signif-25 icant intercept in the model may simply have gone undetected in prior analyses because conventional al-26 lometry assumes implicitly that the intercept is zero; or the intercept may be a spurious finding resulting 27 from bias introduced by the haphazard sampling that underlies "exploratory" analyses like the one reported 28 here. The aforementioned issues can be resolved only by gathering new data specifically intended to ad-29 dress the question of scaling of FMR with body mass in mammals. However, there is no support for the con-30 cept of complex allometry in the relationship between FMR and body size in mammals.

Recent reports indicate that field metabolic rates (FMRs) of mammals conform to a pattern of complex al- 17

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#### 44 1. Introduction

Field metabolic rates (FMRs) of mammals hold the key to under-45standing energy requirements and turnover of animals living under nat-4647ural conditions (Nagy et al., 1999; Speakman, 2000; Nagy, 2001, 2005). Although a variety of factors (like habitat and season) doubtless have 48 some influence on metabolism, most of the variation in FMR among 49species is attributable to variation in body mass (Nagy et al., 1999; 50Speakman, 2000; Nagy, 2005). Early summaries of the scaling of FMR 51with body mass in mammals yielded an allometric exponent of ~0.725 52(Nagy et al., 1999; Anderson and Jetz, 2005; Nagy, 2005), which is tan-53talizingly close to the theoretical value of 0.75 for the scaling of basal 54metabolic rate (Brown and Sibly, 2012) and suggests that the same gen-55eral constraints may apply to FMR that seemingly apply to the basal rate 56(West and Brown, 2005). More recent investigations on the scaling of 5758 FMR in mammals, however, have detected a slight curvilinear pattern (convex) in log-log plots (Capellini et al., 2010; Müller et al., 2012; 59 Hudson et al., 2013; Bueno and López-Urrutia, 2014), thereby pointing 60

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http://dx.doi.org/10.1016/j.cbpa.2016.10.005 1095-6433/© 2016 Published by Elsevier Inc. to a pattern of complex allometry (sensu Strauss, 1993) wherein the ex- 61 ponent in the equation of simple allometry 62

$$y = \mathbf{a} * x^{\mathbf{b}} \tag{1}$$

is an increasing function of body size. Reports of complex allometry 64 have profound implications for the scaling of numerous ecological traits and times (Bueno and López-Urrutia, 2014). 65

It would be premature, however, to attach undue significance to re- 66 ports of complex allometry in FMR of mammals. The prior studies relied 67 on statistical procedures that have changed little since the allometric 68 method was first described in detail by Julian Huxley in his monograph 69 on *Problems of Relative Growth* (Huxley, 1932). Unfortunately, Huxley's 70 method was never adequate for dealing with observations that are cur- 71 vilinear on the logarithmic scale (Reeve, 1940; Reeve and Huxley, 1945; 72 Richards and Kavanagh, 1945), so the appearance of complex allometry 73 may be nothing more than an artifact resulting from inadequate statis- 74 tical analyses. Newer and more versatile analytical procedures now are 75 available, and application of these procedures to the study of mammali-76 an FMRs has the potential to yield insights that were unattainable here-77 tofore (Packard, 2015, 2016). 78

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Here I perform a conventional allometric analysis on a large dataset 79 80 for FMRs of mammals to illustrate the problems and limitations of the traditional method. I then use linear and nonlinear regression, with dif-81 82 ferent assumptions about the nature of random error, to re-examine the same dataset. The regression procedure is a powerful, but underappre-83 ciated, tool for use in allometric research because the computational al-84 85 gorithm is not constrained to fitting a narrow range of power functions 86 based on logarithmic transformations. I show that the perception of 87 complex allometric variation in FMRs of mammals is due entirely to in-88 ability of conventional approaches to fit descriptive equations with explicit, non-zero intercepts and that the exponent in the best regression 89 model is not a varying function of body size. 90

#### 91 **2. Conventional allometric analysis**

Data for FMR and body mass of 116 species of mammal ranging in 92 size from a 7.3 g bat to a 111,400.0 g sea lion were taken from the online 93 94 appendix to the article by Capellini et al. (2010), who compiled records for FMRs that were estimated by the method of doubly labeled water. 95 The observations were transformed to logarithms (base 10) and 96 displayed on a bivariate graph, thereby revealing a slight convex curva-97 ture in the data (Fig. 1A). Equations for a straight line and a quadratic 98 99 polynomial then were fitted to the transformations by ordinary least squares to quantify the relationship between the variables. Both models 100 assumed additive, normal, homoscedastic error on the logarithmic scale 101 (Packard, 2014). Normality of the distributions for residuals was 102assessed with the Shapiro-Wilk test, and constancy of variance was 103 104 evaluated with the Breusch/Pagan test. Akaike's Information Criterion (AIC) was used to compare the models (Burnham and Anderson, 1052002). Generally speaking, models with  $\triangle$ AICs < 4–7 are plausible alter-106 natives to the "best model" ( $\Delta AIC = 0$ ) whereas models with AICs > 14 107have no empirical support (Burnham et al., 2011). 108

109Two potential outliers were identified in a preliminary examination of the full dataset (Fig. 1A). Although neither of these outliers was par-110ticularly influential (as judged by Cook's Distance), they were removed 111 from the dataset before proceeding further (removing these observa-112 tions had no substantive effect on the outcome). The model for the 113 114 straight line fitted to observations for the remaining 114 species explains >95% of the variation in log FMR (Table 1). However, the fitted 115 equation does not perform well at the upper and lower ends of the 116 size range, where observations for the response are appreciably higher 117 118 than expected for a rectilinear model (Fig. 1A). This departure from linearity is confirmed by the curvature in a plot of standardized residuals 119 against fitted values (Fig. 1B). The rectilinear fit also is heteroscedastic 120 121 (Table 1), owing to greater variability in the response for larger species (Fig. 1B). Back-transforming the deterministic equation is unlikely to 122123yield a good description for the pattern of variation in the original data because the fit to logs failed to satisfy the fundamental requirement 124for linearity (Reeve, 1940; Reeve and Huxley, 1945; Richards and 125Kavanagh, 1945). 126

The quadratic equation explains only 1.3% more of the variation in 127128log FMR than is explained by the rectilinear fit (Table 1), but the mean 129function for the quadratic successfully captures the curvature in the observations (Fig. 1A) and the model is strongly favored by AIC (Table 1). 130Although the quadratic seemingly violates the assumption of homoge-131neity of variance (Table 1), the plot of standardized residuals indicates 132that the departure from homoscedasticity is trivial and that residuals 133 are satisfactory (Fig. 1C). The quadratic model consequently is accepted 134 here as a reasonable fit to the observations. 135

How should the quadratic model fitted to logarithms be interpreted?
Three problems with the model are major concerns: the first problem is
one of general importance to the entire field of biology, namely, the use
of quadratic polynomials to describe biological data; the second problem concerns the use of quadratic equations in traditional allometry;
and the third problem focuses on the present dataset for FMRs of
mammals.



**Fig. 1.** (A) Rectilinear and quadratic models fitted to logarithmic transformations of data for FMR and body mass of 114 species of mammal. Stars identify possible outliers that were identified in a preliminary examination of the data (Studentized deleted residuals = -3.8 for *Zyzomys argurus* and +3.6 for *Arctocephalus gazella* in the fit of the quadratic model to the full dataset). The outliers were not included in the final analysis. (B) Standardized residuals vs. fitted values from the fit of the straight line to logarithmic transformations. Note the bowl-shaped pattern confirming that transformation field to linearize the distribution. (C) Standardized residuals vs. fitted values from the fit of the quadratic polynomial to logarithmic transformations.

First, statisticians have tried for years to convince biologists not to 143 resort to quadratic equations (or other polynomials) to describe bivari- 144 ate data of any kind (e.g., Finney, 1989; Paine et al., 2012). A quadratic 145 equation traces the path of a parabola (http://www.mathopenref.com/ 146 quadraticexplorer.html), so a quadratic seldom (if ever) is a realistic 147 representation of pattern in biological data. The fitted model may follow 148 the path of observations quite well over the limited range for the 149

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