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## Invited reply

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## Evolutionary biology

## Response to Packard: make sure we do not throw out the biological baby with the statistical bath water when performing allometric analyses

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Quantifying accurately the relationship between a phenotypic trait and body mass is a long-standing challenge in evolutionary biology and constitutes the core of allometric analyses. The traditional approach to study allometric relationships (expressed as  $y = b x^a$ , [1]) is to fit a linear (or more recently, a quadratic [2]) model on log-transformed data. This allows estimating  $a$ , the allometric exponent which corresponds to the growth ratio between  $y$  and  $x$ , and  $b$  the allometric intercept that depends on the initial values of  $y$  and  $x$ , respectively [1]. Further, the comparison between linear and quadratic models allows the testing for possible changes in the allometric coefficient with increasing size. Using such a standard approach, we have recently shown [3] that the allometric exponent of antler size decreases with increasing body mass in extant cervids larger than 110–120 kg. Our best models, a threshold model and a quadratic model fitted the data particularly well. However, Packard [4] qualified our statistical approach of ‘unnecessary and incorrect’ and cast doubt on our take-home message that the allometric exponent of antler size changes with body mass.

Packard’s criticism falls into the recent debate on using logarithmic transformations in allometric studies (e.g. [5,6]). Packard (e.g. [6,7]) has repeatedly argued that logarithmic transformations are often unjustified, do not meet assumptions of linear models and are misleading. Instead, this author advises to fit a set of power functions on raw data. The expected type of biological variation is a cornerstone in the logarithmic versus arithmetic scale debate [8]. Because it is most probable that biological errors are generated by multiplicative processes similar to those responsible for growth, log-transformations reliably control for this size-dependent variability, and ensure the required normal distribution of the residuals. Fitting a power function assumes that errors are additive rather than multiplicative [8]. Alternatively, Packard [9] argues that the ‘multiplicative nature’ of any biological trait can be taken into account by controlling for heteroscedasticity when fitting generalized nonlinear models. Using a log–log scale model or a power model with heteroscedastic error should provide similar results. To check whether this is the case, we consider the simple situation of a constant allometric exponent with mass illustrated by the antler size allometry across deer species below 113 kg ( $n = 22$ ). As expected, both estimates of allometric exponent are close ( $1.326 \pm 0.152$  for the log–log scale model versus  $1.235 \pm 0.108$  for the power model with heteroscedastic error), leading to the same biological interpretation of a strong positive allometry (the expected exponent for isometry between a mass and a linear measurement being one-third). So, how can we explain the discrepancy observed between the two methods when analysing the whole deer dataset?

To get a model satisfactorily fitting the data, Packard [4] used a three-parameter power function with additive heteroscedastic error ( $\hat{y} = -40.536 + 16.414x^{0.399}$ ). This model hinders any biological interpretation of the parameters. In particular, the exponent in this model does not inform anymore about the value or the change of the allometric exponent (*sensu* Huxley [1]), rendering this model less easily applied in the context of allometric studies. Moreover, the biological meaning of the value  $-40.536$  is unclear. This contrasts with the parameters of a quadratic model  $\log(y) = \log(b) + a\log(x) + c(\log(x))^2$  that are relatively straightforward to interpret (e.g.  $a$ : the maximum allometric exponent occurring at very low  $x$ , and  $c$  the strength of the reduction in the allometric exponent with increasing  $x$ ). This additional parameter leads Packard's model to predict that cervids below 10 kg should not have any antlers, which is not supported by empirical evidence. Indeed, male Northern pudu (*Pudu*

*mephistophiles*) weigh much less than 10 kg but wear antlers [10].

We certainly agree that the three-parameter power model with heteroscedastic error proposed by Packard provides the best fitting line and thus has some merit in a pure statistical sense. His parsimonious model offers a reliable way to describe the allometric relationship. However, allometric analyses do not aim only at fitting lines, and biological inferences are generally looked for, although often overlooked [5]. When it comes to deciphering the complex biological processes underlying the strength and the shape of allometric relationships, the traditional 'log-transformed' methodology should be preferred to Packard's approach because this latter is difficult to interpret in biological terms [11].

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