



Full Length Article

Application of Linear, Quadratic and Cubic Regression Models to Predict Body Weight from Different Body Measurements in Domestic Cats

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ABSTRACT

The aims of this study were to predict body weight (BW) from different body measurements and to determine the best regression model for domestic cats. For this aims, a total of 48 adult Turkish cats (20 females & 8 males Turkish Angora; 13 females & 7 males Turkish Van) were used. In the study, wither height (WH), body length (BL) and head circumference (HC) were assumed as independent variables, whereas body weight was used as dependent variable. Linear, quadratic and cubic effects of the independent variables were included in the assumed model as $Y = b_0 + b_1X + b_2X^2 + b_3X^3 + e$. Where Y = body weight; b_0 = the intercept; X = independent variables, (WH, BL, or HC); b_1 , b_2 and b_3 = regression coefficients and e = random error. Conceptual predictive (Cp) and Akaike information criterion (AIC) were used to determine the most suitable model among the assumed models. The model that has the smallest Cp and AIC values is the best model. The R^2 values from the regression indicate the BL ($R^2 = 0.50$) to be moderately related to the BW. Neither the quadratic term nor the cubic term was significant for all body traits, whereas the linear term was highly significant ($p < 0.001$) for all independent variables. Since the maximum number of independent variables is three, there were seven possible different models. It can be concluded that cat body weight was explained with the following model. (BW) = - 4.53 + 0.11 WH + 0.13 BL with p- values, <0.001, 0.0083, and <0.001 for the intercept, b_1 and b_2 , respectively with $R^2 = 0.57$. © 2011 Friends Science Publishers

Key Words: Cat; Regression analysis; Body weight estimation

INTRODUCTION

In a domestic cat population, body weight is an important trait that is used in evaluating body condition (Erat & Arikan, 2010) and health status (Lund *et al.*, 2005) of cats, in computing dosages and in prescribing drugs. Body weight and condition score are also often used for assessing nutritional condition of dog and cat (Laflamme, 1997; Esfandiari & Youssefi, 2010).

A lot of techniques, which are simple or sophisticated and expensive or inexpensive, are available to get information on animal's body traits. The easiest way to assess an animal's body mass is to weigh the animal. However, under some situations scale may not be available and prediction of body weight from body measurements could be preferred practically (Latshaw & Bishop, 2001).

Multiple regression analysis has been used widely to describe quantitative association between dependent (body weight) and independent variables (hearth girth, body length & wither height etc.) in animal studies (Cankaya, 2009). Several studies on cattle, sheep and goat (Mohammed & Amin, 1997; Wilson *et al.*, 1997; Atta & El khidir, 2004;

Topal & Macit, 2004; Adeyinka & Mohammed, 2006; Bagui & Valdez, 2007), dog and cat (Pendergrass *et al.*, 1983; Valdez & Recuenco, 2003), horse and donkey (Pearson & Ouassat, 1996; Marante *et al.*, 2007) and poultry (Latshaw & Bishop, 2001; Grona *et al.*, 2009) have been conducted to predict body weight from body measurements.

The body weight estimation from using external body measurements on domestic cats is scarce. Valdez and Recuenco (2003) estimated the body weight of Philippine domestic cats using some external body measurements. Therefore, the present study was designed to predict body weight of Turkish cats (*Felis catus*) from wither height, body length and head circumference and to choose the most appropriate regression model.

MATERIALS AND METHODS

Source of data: A total of 48 adult Turkish cats (20 females & 8 males Turkish Angora; 13 females & 7 males Turkish Van) were used (Fig. 1). All cats were solid white in color and sexually intact. Averages of weight and age for these cats were 3.39 kg and 2.71 years, respectively.

Measurements: Live body weight (BW), wither height

(WH), body length (BL) and head circumference (HC) were measured. The measurements of the body traits were described in Erat and Arikan (2010). The BW was taken with a scale of 5 g of precision. A cloth tape was used for measuring the WH, BL and HC. The WH was the distance from the floor beneath the cat to the top of withers. The BL was the distance from the point of the shoulder to the ischiatic tuberosity. The HC was measured by wrapping the cloth tape around the circumference of the cat's head just behind the ears and across the forehead. The BW was recorded in kilograms and the other measurements were recorded in centimeters. All measurements were taken from the left side of the cats.

Statistical analysis: In the present study, the WH, BL and HC measurements were used as independent variables, while the BW was considered as a dependent variable. Linear, quadratic and cubic effects of the independent variables were considered as shown in the following model (Heinrichs *et al.*, 1992):

$$Y = b_0 + b_1X + b_2X^2 + b_3X^3 + e$$

Where Y = body weight; b_0 = the intercept; X = independent variables, either the WH, BL, or HC; b_1 , b_2 and b_3 = regression coefficients; and e = random error. Conceptual predictive (Cp) and Akaike information criterion (AIC) (Kaps & Lamberson, 2004) were used to determine the most suitable model among the assumed models.

The Cp and AIC statistics were defined by (Kaps & Lamberson, 2004) as:

$$Cp = p + \frac{(MS_{RES} - \hat{\sigma}_0^2)(n - p)}{\hat{\sigma}_0^2}$$

Where: MS_{RES} = residual mean square for the candidate model.

$\hat{\sigma}_0^2$ = variance estimate of the true model.

n = the number of observations.

p = the number of parameters of the candidate model (the number of independent variables + 1).

Usually, the estimate of variance from the full model, which is the model with the maximal number of parameters, is used. Then:

$$\hat{\sigma}_0^2 \cong MS_{RES-FULL}$$

$$AIC = n \log(SS_{RES} / n) + 2p$$

Where: SS_{RES} = Residual sum square, n and p are explained above.

Phenotypic correlations between the BW and the external body parameters were also calculated. Statistical analyses were performed using MEANS, REG and CORR procedures of SAS v.8.2 (The SAS Institute Inc., Cary, NC) statistical package program.

Table I: Descriptive statistics for body traits of the Turkish cats

Body trait	N	Mean	SE ^a	Min.	Max.	CV ^b (%)
Body weight (kg)	48	3.39	0.11	1.81	5.52	21.92
Wither height (cm)	48	26.31	0.29	22.00	31.00	7.73
Body length (cm)	48	37.06	0.45	29.00	44.00	8.39
Head circumference (cm)	48	25.06	0.28	21.00	29.00	7.87

^a Standard Error, ^b Coefficient of Variation

Fig. 1: Turkish Angora (left) and Turkish Van (right) cats. Photos were taken by Serkan Erat



RESULTS AND DISCUSSION

Descriptive statistics for the body traits of Turkish cats are given in Table I. The BW showed the highest variation among the traits measured. Erat and Arikan (2010) reported that similar averages of these body measurements were 3.64 kg for the BW, 26.93 cm for the WH, 37.21 cm for the BL and 26.09 cm for the HC, for both Turkish cats. Hendriks *et al.* (1997) reported an average body weight of 3.80 kg for both gender of intact cats and Allan *et al.* (2000) reported an average body length of 37.00 cm for their cats. These values were in agreement with the averages of body weight and length of Turkish cats in the present study.

Table II presents regression analysis results of cat body weight on different body measurement using individual observations. The R^2 values from the regression analysis indicated that the BL ($R^2=0.50$) was connected in medium level with the BW. The quadratic term and the cubic term were insignificant for all body traits, whereas the linear term was highly significant ($P < 0.001$) for all independent variables. Several studies, conducted to predict body weight for different animal species, reflected that the heart girth was one of the greatest body weight predictors (Heinrichs *et al.*, 1992; Pearson & Ouassat, 1996; Wilson *et al.*, 1997; Gueye *et al.*, 1998; Valdez & Recuenco, 2003; Atta & El khidir, 2004; Adeyinka & Mohammed, 2006). However, Aziz and Sharaby (1993) reported that height at withers could be considered as a better estimator for body weight than hearth girth. Wickersham and Schultz (1963) reported that wither height could be one of the best skeletal measurements in some instances since it is not influenced by body condition. Sulieman *et al.* (1990) also reported that body length and wither height were less variable than hearth girth, because hearth girth was the most variable measurement influenced by body condition and by physiological status of the animal in some instances.

Table II: Regressions of the Turkish cats' body weight on various body measurements using individual observations

Measurement	Intercept	Linear	Quadratic	Cubic	(R ²) ^a	Model p-values
Wither height	- 2.08	0.20799***			0.32	< 0.001
	- 7.83	0.64580	- 0.00828		0.33	< 0.001
	- 147.15	16.59213	- 0.61377	0.00763	0.34	< 0.001
Body length	- 2.86**	0.16878***			0.50	< 0.001
	- 1.14	0.07507	0.00127		0.50	< 0.001
	- 60.71	5.02473	- 0.13478	0.00124	0.51	< 0.001
Head circumference	- 1.81	0.20757***			0.30	< 0.001
	-12.18	1.04404	- 0.01677		0.31	< 0.001
	- 92.87	10.82266	- 0.40967	0.00524	0.32	< 0.001

P<0.01, *P < 0.001

^a R² is the coefficient of determination which gives the proportion of variability in Turkish cat body weight (Y) explained by body measurements (Xs)

Table III: Conceptual predictive (Cp) and Akaike information criterion (AIC) for selecting an optimal model for predicting body weights of the Turkish cats

Number of independent variables in Model	p ^a	df ^b	Cp	AIC	R ²	SSE ^c	Variable in Model ^d
2	3	45	3.0980	- 63.9419	0.5705	11.17945	WH BL
3	4	44	4.0000	- 63.1250	0.5809	10.90727	WH BL HC
2	3	45	7.6375	- 59.3383	0.5272	12.30477	BL HC
1	2	46	8.7279	- 58.4392	0.4978	13.07086	BL
2	3	45	18.2044	- 50.0743	0.4266	14.92422	WH HC
1	2	46	27.0831	- 44.1013	0.3230	17.62097	WH
1	2	46	29.2176	- 42.6812	0.3026	18.15009	HC

^a The number of parameters of the candidate model. ^b degree of freedom. ^c residual sum square

^d WH=Wither height, BL= Body length, HC= Head circumference

Therefore, linear model with wither height and body length compared to the models in Table III could be better skeletal measurements for predicting body weight as in the present study. Additionally, head circumference seems first to be accounted for predicting body weight.

In the present study, results of seven possible different models are depicted in Table III. The Cp and AIC were used to detect the most suitable model (Table III). The value of Cp for the model with the WH and BL was found to be better than other models in Table III. Also, there was a bit better in R² for models with the WH, BL, HC compared to the model with the WH and BL. The AIC value was also the smallest for the model with the WH and BL. Results revealed that the model with the WH and BL could explain cat body weight better than other models (Kaps & Lamberson, 2004). The parameter estimates were b₀ = - 4.53, b₁ = 0.11 and b₂=0.13, for the intercept, the WH, and BL, respectively. So the model became; cat body weight (BW) = - 4.53 + 0.11 WH + 0.13 BL with p- values, <0.001, 0.0083 and <0.001 for the intercept, b₁ and b₂, respectively with R² = 0.57. Aziz and Sharaby (1993) also used Cp statistics for comparing efficiency of different models in addition to determination coefficient (R²) and adjusted R². They found that heart girth may be better for predicting body weight of Najdi sheep than wither height based on R² and adjusted R². However, they concluded that wither height was a better predictor than hearth girth for predicting body weight of Najdi sheep based on Cp statistic. Neter *et al.* (1996) reported that the bias of the regression model was small when subsets of independent variables with small Cp values had a small total mean squared error and Cp value

Table IV: Phenotypic correlations between the body measurements of the Turkish cats

	WH	BL	HC
BW	0.57***	0.71***	0.55***
WH		0.47**	0.47**
BL			0.58***

P<0.01, *P<0.001

WH=Wither height, BL= Body length, HC= Head circumference

was near p. AIC was also used for model selection by others (Akaike, 1987; Bozdogan, 1987). Bozdogan (1987) described AIC as a simple and versatile procedure and reported that AIC had provided a new and modern way of thinking on how to solve many important statistical modeling problems. Beal (2005) showed that AIC for multivariate model selection was superior to heuristic methods such as forward selection, backward elimination, stepwise regression and minimizing Root of Mean Square Error (RMSE) using simulated data with a known underlying model. Valdez and Recuenco (2003) found that addition of second and third variable to regression equations slightly increased R² values but their R² values were never below 0.70.

Phenotypic correlations of the BW with the BL, WH and HC traits were also calculated (Table IV). It was found that the BW showed the highest correlation coefficient value with the BL (r=0.71; p<0.001) followed by the correlation coefficient values with the WH (r=0.57; p<0.001) and the HC (r=0.55; p<0.001). Similarly, positive correlation coefficients of body weight with external body measurements were also reported for mongrel dogs (0.76 for length: weight; 0.85 for circumference: weight & 0.41 for

height: weight) (Pendergrass *et al.*, 1983), Philippine domestic cats (Valdez & Recuenco, 2003) and for Turkish cats (Erat & Arikan, 2010).

CONCLUSION

Body length was the highly correlated with body weight and combining body length with other body measurements (with height & head circumference) in multiple regression produced higher R^2 . C_p and AIC values could be used to choose the best suitable model among the assumed models. Despite shortcomings of the cat data, an attempt was made to predict the body weight of the cats from different body measurements. Therefore, the formulas found in the present study could be helpful for an approximate estimation of cat body weight with caution since R^2 values were either low or medium level. Others factors such as gender and age of a cat may also affect body weight estimation. However, for convenience of users who work under field conditions, prediction equations can be constructed without considering those effects.

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