



Adjusting weight growth curve of male quails *Coturnix japonica* reared in the semi-arid region of the State of Pernambuco

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ABSTRACT. This study adjusted different regression models to describe the growth pattern of meat quails from birth to 42 days of age. Data of 300 male quails were used. Weight and height information of all quails were collected weekly from the 1st to the 42nd day of age. Body weight of poultry was subjected to the polynomial, logistic, Gompertz, Weibull, and log-normal regression models. The criteria used to choose the best model to explain the growth curve of quails were the coefficient of determination of the model, Akaike's information criterion, sum of squared residuals and Willmott's index. For all the models used, the variables age and height were significant to explain the weight of quails. The polynomial ($R^2 = 99.99\%$, AIC = 24.68, SSR = 27.5, $d = 0.9999$) and log-normal ($R^2 = 99.60\%$, AIC = -17.5, SSR = 107.15, $d = 0.9989$) models presented the best fit criteria and were recommended to explain the growth of quails.

Keywords: modelling; poultry; regression model.

Received on April 23, 2018.

Accepted on June 14, 2018.

Introduction

Coturniculture is a branch of poultry farming that has been highlighting in recent years (Drumond et al., 2013). The quail is still little studied resulting in a great shortage of information regarding genetics, nutrition and management. The main phenotypic characteristic of commercial interest in meat quail farming is body weight, which has medium to high heritability mainly during the growth phase (Knížetová, Hyánek, Hyankova, & Bělíček, 1995). Japanese quails are excellent laying, but males have been studied for meat production (Raji, Mbap, & Aliyu, 2014; Rocha-Silva et al., 2016).

The curve that describes a sequence of measurements of a particular characteristic of a species or individual as a function of time, usually weight, height, diameter, length is called growth curve. This type of curve generally presents exponential or sigmoidal growth form (S- shaped) (Lucena, 2017). The study of growth by means of fits of a function that describes a period of life of animal, relating weight and age, has been researched by several authors (Macedo et al., 2014; Morais et al., 2015; Teixeira et al., 2012b). Data analyses of repeated measures are of fundamental importance in animal production and are analyzed over several conditions such as: weather, climate and air humidity (Freitas, 2005). Several models (Brody, Logistic, Gompertz, Von Bertalanffy, Gamma and cubic polynomial) have been proposed to explain the biological growth of animals (Lucena, 2017; Lucena, Holanda, Holanda, & Sousa, 2017; Nascimento, Ribeiro, Rocha, & Lucena, 2017; Souza et al., 2011).

In poultry farming, stands out the studies of Lucena et al. (2017), Michalczyk, Damaziak, and Goryl (2016) and Selvaggi, Laudadio, Dario, and Tufarelli (2015) observed that the logistic, Gompertz and Richard models were most adequate to explain the weight growth of broiler poultry. Liu, Li, Li, and Lu (2015), Zhao et al. (2015), Al-Samarai (2015) and Mohammed (2015) suggested the Gompertz and logistic models best adequate to explain the weight of poultry. Eleroğlu, Yıldırım, Şekeroğlu, Çoksöyler, and Duman (2014) verified that logistic model was more suitable to explain the weight of poultry reared in organic systems. On the other hand, in coturniculture, stands out the works of Narinc, Karaman, Firat, and Aksoy (2010), Mota et al. (2015) and Rocha-Silva et al. (2016), in which

they verified that Gompertz model was indicated as the most suitable to explain the weight growth of quails. Mota et al. (2015) observed that the logistic model was the most weight growth to explain the weight of quails. Raji et al. (2014) observed that the model most suited to explain the weight growth of quails was Weibull. Drumond et al. (2013) verified that the logistic and Gompertz models presented the best results to explain the growth of male and female quails. Bonafé et al. (2011) and Teixeira et al. (2012a) verified that the weight growth of quails can be explained by Legendre polynomial models.

The present study aimed to evaluate the weight growth of quails from 1 to 42 days of age by fitting regression models.

Material and methods

The research was conducted from November 1st to December 13rd, 2017 by the Group of Studies on Pig and Poultry (GESA) of Federal Rural University of Pernambuco (UFRPE), Academic Unit of Serra Talhada (UAST), located in the municipality of Serra Talhada, Microregion of Sertão do Pajeú, Meso-region of Sertão Pernambucano.

The study sample is composed of 300 male Japanese quail. Seven evaluations (1, 7, 14, 21, 28, 35 and 42 days after the birth of quails) of weight and height were performed as shown in Figure 1.

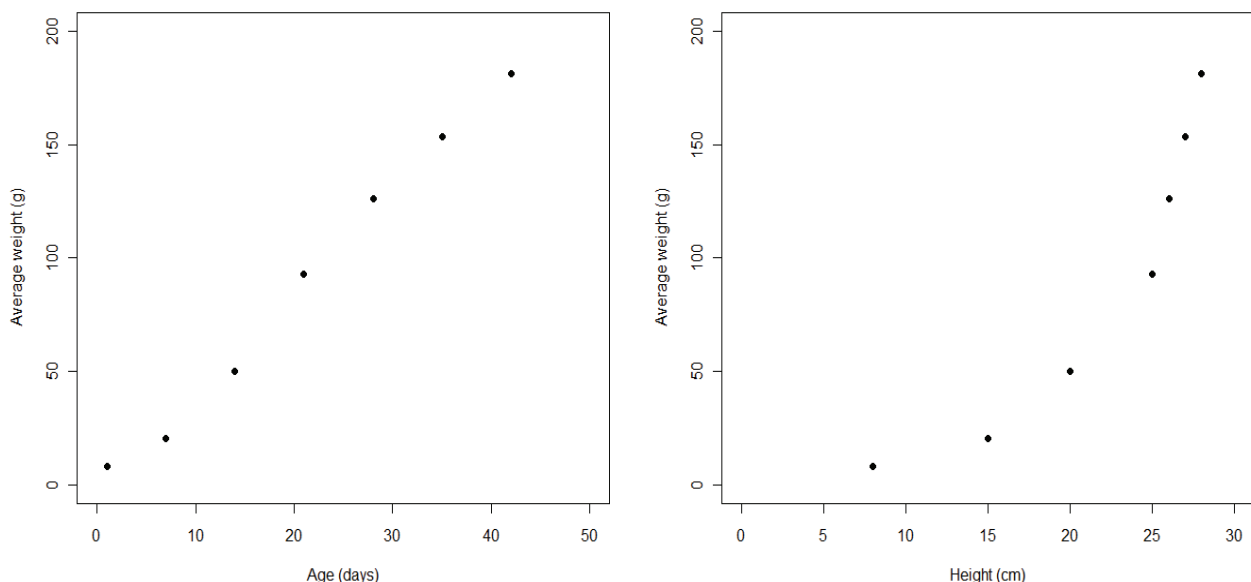


Figure 1. Average weight of quails in relation to their age (a) and height (b).

At birth, the poultry were individually identified, weighed and randomly transferred to the boxes. Quails were housed in ten boxes (replicates) of dimensions 2.00 x 2.00 m, with 30 birds per box. They were reared on concrete floor with shaving litter, and artificial heating by with 250W infrared lamps until 21 days of age. The quails were given water and diet at will throughout the experimental period.

The quails received one type of diet, one up to 42 days, a diet with an average of 22% crude protein (CP) and 2900 kcal kg⁻¹ metabolizable energy (ME), according to recommendations of nutritional requirements, Table 1.

The regression models listed in Table 2 were initially proposed to explain the average weight of quails according to their average lifetime and height.

The models were evaluated by the following criteria: coefficient of determination of model (R^2), Akaike's information criterion (AIC), by sum of squared residuals (SSR) and by Willmott's index (d). Be \hat{Y}_i the value of i -th weight after fitting of model, then the sum of squared residuals for this study is defined by the following expression:

$$SSR = \sum_{i=1}^n (Y_i - \hat{Y}_i)^2$$

Table 1. Diet supplied to quails from to 42 days of age.

Ingredients	Food Composition
	1 to 42 days
Corn	57.3354
Soybean meal (46%)	37.9859
Limestone	1.1569
Bicalcium phosphate	1.4509
Salt	0.3475
DL-methionine (99%)	0.1671
Soy oil	1.3140
Vitini-birds ¹	0.1000
Min-birds ²	0.1000
L-Threonine (98%)	0.0403
Total	100.00
Calculated composition	
CP (%)	22.0000
ME (Mcal/kg)	2.900
Calcium (%)	0.900
Disp. phosphorus (%)	0.375
Crude Fiber (%)	3.366
Digestible Arginine birds (%)	1.422
Digestible Histidine (%)	0.549
Digestible Isoleucine (%)	0.878
Digestible Leucine (%)	1.774
Digestible Lysine (%)	1.180
Digestible Met+Cis (%)	0.775
Digestible Methionine birds (%)	0.480
Digestible Phenylalanine (%)	1.020
Digestible Phenyl + Tyrosine (%)	1.719
Digestible Threonine (%)	0.790
Digestible Tryptophan (%)	0.254
Digestible Valine (%)	0.921
Ether Extract (%)	3.800
Sodium (%)	0.176

¹Vitamin supplement per kg of product: vit. A 3,750,000UI; vit. D3 750,000UI; vit. E 7,500mg; vit. K 3 1,000mg; vit.B1 750mg; vit.B2 1,500mg; vit. B6 1,500mg; vit. B12, 7,500mcg; vit.C 12,500mg; biotin 30 mg; niacin 10,000mg; folic acid 375; pantothenic acid 3,750mg; hill 10.000mg; methionine 400,000mg. ²Supplement mineral per kg of product: selenium 45mg; iodine 175mg; iron 12,525mg; copper 2,500mg; manganese 19,500mg; zinc 13,750mg; prom. Prod 15,000mg; coccidiostat 10,000mg; antioxidant (BHT) 500mg.

Table 2. Regression models to explain the average weight of quails.

Regression models	Equation
Polynomial	$Y_i = \beta_0 + \beta_1 T_i + \beta_2 T_i^2 + \beta_3 A_i + \beta_4 A_i^2 + \beta_5 A_i T_i + \varepsilon_i$
Gompertz	$Y_i = w \exp(-\exp(\beta_0 + \beta_1 T_i + \beta_2 T_i^2 + \beta_3 A_i + \beta_4 A_i^2 + \beta_5 A_i T_i)) + \varepsilon_i$
Logistic	$Y_i = \frac{w}{1 + \exp(\beta_0 + \beta_1 T_i + \beta_2 T_i^2 + \beta_3 A_i + \beta_4 A_i^2 + \beta_5 A_i T_i)} + \varepsilon_i$
Weibull	$Y_i = \exp(-\beta_0 + \beta_1 T_i + \beta_2 T_i^2 + \beta_3 A_i + \beta_4 A_i^2 + \beta_5 A_i T_i) + \varepsilon_i$
Log-normal	$Y_i = \exp(-\beta_0 + \beta_1 T_i + \beta_2 T_i^2 + \beta_3 A_i + \beta_4 A_i^2 + \beta_5 A_i T_i) + \varepsilon_i$

Where, Y_i is the average weight of i -th quails; T_i is the i -th age of quails; A_i is the i -th average height of quails; AT_i is the i -th iteration age and height of quails and ε_i is the i -th error associated with weight of quails, in which, ε_i presents normal distribution of mean 0 and constant variance $\sigma^2 > 0$ for the models: polynomial, Gompertz; logistic and log-normal, while for the Weibull model ε_i presents Weibull distribution of parameters α and β . The incognitos w (maximum growth weight of quail), β_0 (initial weight of quail); β_1 (growth rate of quail in relation to the age); β_2 (growth rate of quail in relation to the square age); β_3 (growth rate of quail in relation to the height); β_4 (growth rate of quail in relation to the square height) and β_5 (growth rate of quail in relation to the interaction between age and height) are the parameters associated with the models.

The coefficient of determination of the model (R^2) is expressed by the ratio between the sum of square of model (SSM) and sum of square of total (SST), that is,

$$R^2 = \frac{SSM}{SST} = 1 - \frac{SSR}{SST} = 1 - \frac{\sum_{i=1}^n (Y_i - \hat{Y}_i)^2}{\sum_{i=1}^n (Y_i - \bar{Y}_i)^2}$$

The Akaike's information criterion (AIC) defined by Akaike (1974) is given by:

$$AIC = -2 \ln L(x \setminus \hat{\theta}) + 2(p)$$

where, $L(x \setminus \hat{\theta})$ is the maximum likelihood function, defined as the product of density function and p is the number of model parameters.

The index d defined by Willmott (1981) is given by:

$$d = 1 - \frac{\sum_{i=1}^n (\hat{Y}_i - Y_i)^2}{\sum_{i=1}^n (|\hat{Y}_i - \bar{Y}| + |Y_i - \bar{Y}|)^2}$$

where, \bar{Y} is the mean weight of quails (Y_i).

Results and discussion

Figure 2 showed that logistic model was very accurate up to the 35 days of age of quails, at 42 days, the model began to underestimate the weight of quail. The logistic model presented a high explanatory power ($R^2 = 99.96\%$), low AIC = -18.78 and high Willmott index ($d = 0.9981$, Table 3). Similar results were reported by Mota et al. (2015) studying laying and meat quails, males and females, of different genetic groups found a R^2 higher than 95% and AIC higher than 270.

In the evaluation of Gompertz regression model, we verified that the model did not present good adjustments in relation to the weights observed, Figure 3. The Gompertz model presented $R^2 = 99.81\%$, AIC = -15.48, SSR = 215.61 and $d = 0.9979$ (Table 3). Drumond et al. (2013) studying female meat quails fed diets containing 25% crude protein and 2900 kcal metabolizable energy and found R^2 higher than 97.61%, Rocha-Silva et al. (2016) evaluating meat quails of the LF1 lineage verified R^2 higher than 92% and AIC higher than 132; while Narinc et al. (2010) evaluating Japanese quails fed a diet containing 24% crude protein and 3000 kcal metabolizable energy verified $R^2 = 99.99\%$ and AIC = -2.01. These results corroborate the findings of this research.

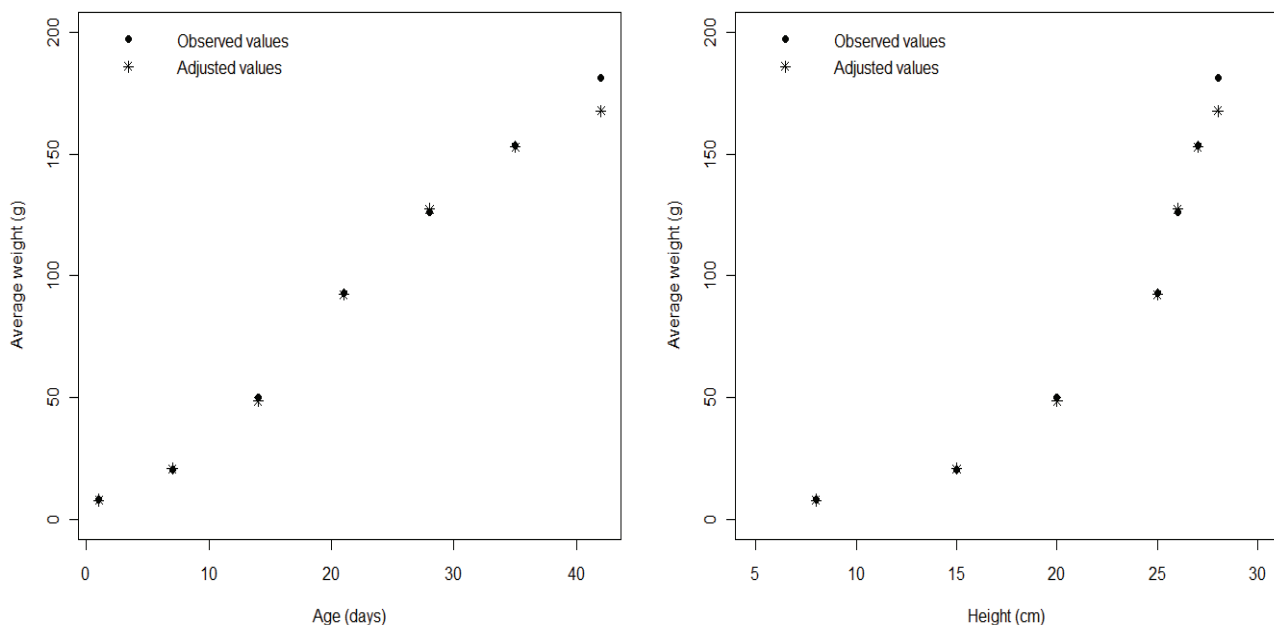


Figure 2. Fit of logistic model.

Table 3. Equation and model evaluated of criteria (R^2 - coefficient of determination; AIC - Akaike information criterion; SSR - sum of squared of residuals; d - Willmott's index).

Regression Models	Equation	Goodness of fit criteria			
		R^2	AIC	SSR	d
Polynomial	$Y_i = -2.73T_i + 1.96A_i - 0.088T_i^2 - 0.132A_i^2 + 0.425AT_i$	99.99	24.68	27.5	0.9999
Gompertz	$Y_i = 181.25 \exp^{-\exp(0.949 - 0.109T_i + 0.039A_i)}$	99.81	-15.48	215.61	0.9979
Logistic	$Y_i = \frac{181.25}{1 + \exp(3.622 - 0.11T_i - 0.054A_i)}$	99.96	-18.78	190.14	0.9981
Weibull	$Y_i = \exp(0.927 + 0.007T_i + 0.143A_i)$	99.39	-17.20	164.07	0.9985
Log-normal	$Y_i = \exp(0.971 + 0.011T_i + 0.135A_i)$	99.60	-17.50	107.15	0.9989

The Weibull regression model did not present good adjustments in relation to the weight of quails (Figure 4). The Weibull regression model presented a high coefficient of determination ($R^2 = 99.39\%$) and

index $d = 0.9985$, low AIC = -17.20 and smaller SSR = 164.07 compared to logistic and Gompertz models (Table 3). Raji et al. (2014), studying Japanese quails fed a diet containing 23% crude protein and 3000 kcal metabolizable energy, observed that the Weibull model presented a R^2 value of 99.9% and AIC higher than 471.

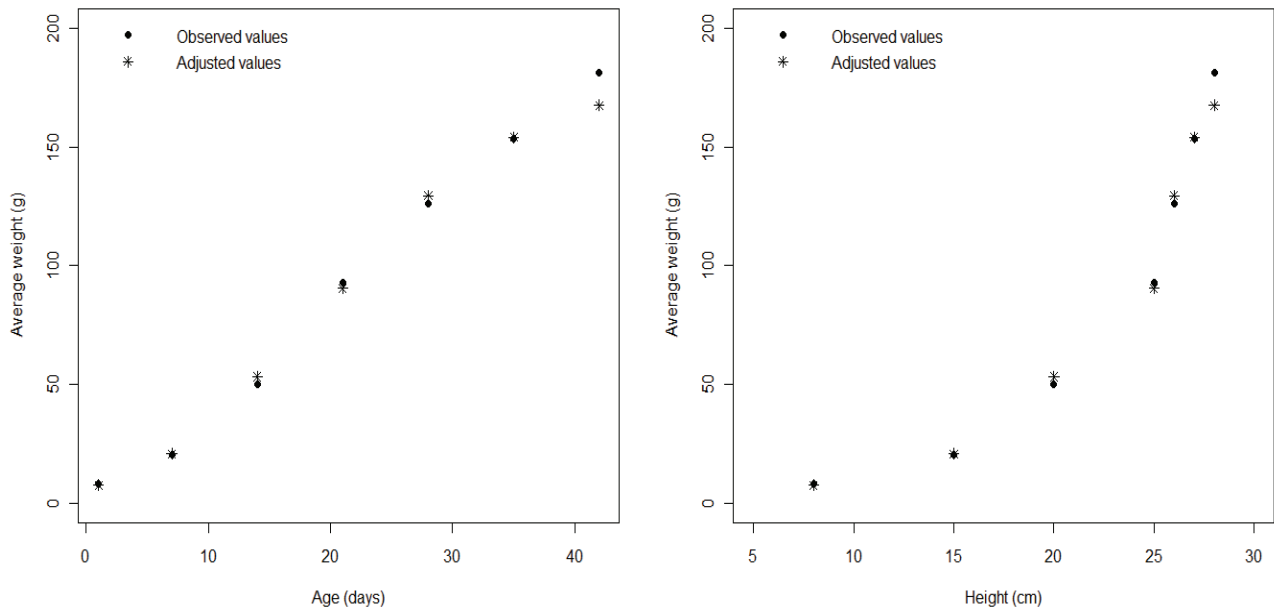


Figure 3. Fit of Gompertz model.

The log-normal regression model showed good adjustments of quail weight when compared to observed values of quail weight throughout the evaluated period (Figure 5). The log-normal model presented a high $R^2 = 99.60\%$, low AIC = -17.50, SSR = 107.15 (lower than the logistic, Gompertz and Weibull models) and index $d = 0.9989$ (larger than the logistic, Gompertz and Weibull models, Table 3).

The polynomial regression model presented similar adjustments to observed weights of quails for all evaluation periods (Figure 6). The polynomial regression model presented high explanatory power ($R^2 = 99.99\%$), low AIC = -24.68 and SSR = 2.5 and greater Willmott index ($d = 0.9999$) than the other models evaluated (Table 3). Similar results were found by Bonafé et al. (2011) and Teixeira et al. (2012a) who studied mating of cut quails belonging to the genetic groups UFV1 and UFV2.

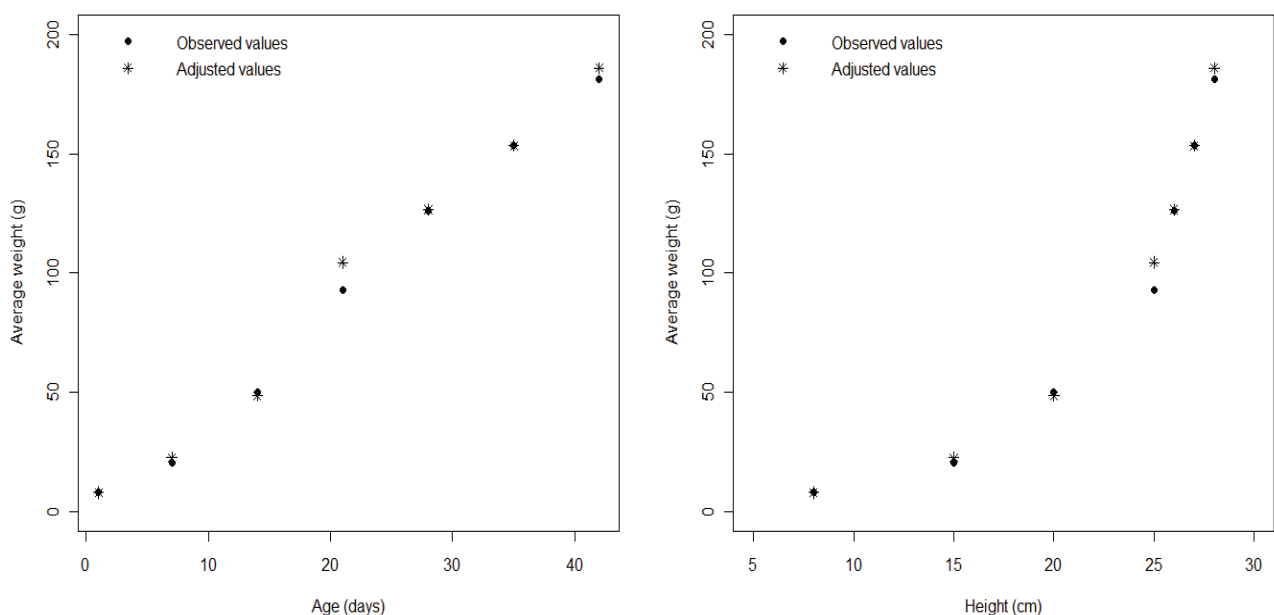


Figure 4. Fit of Weibull model.

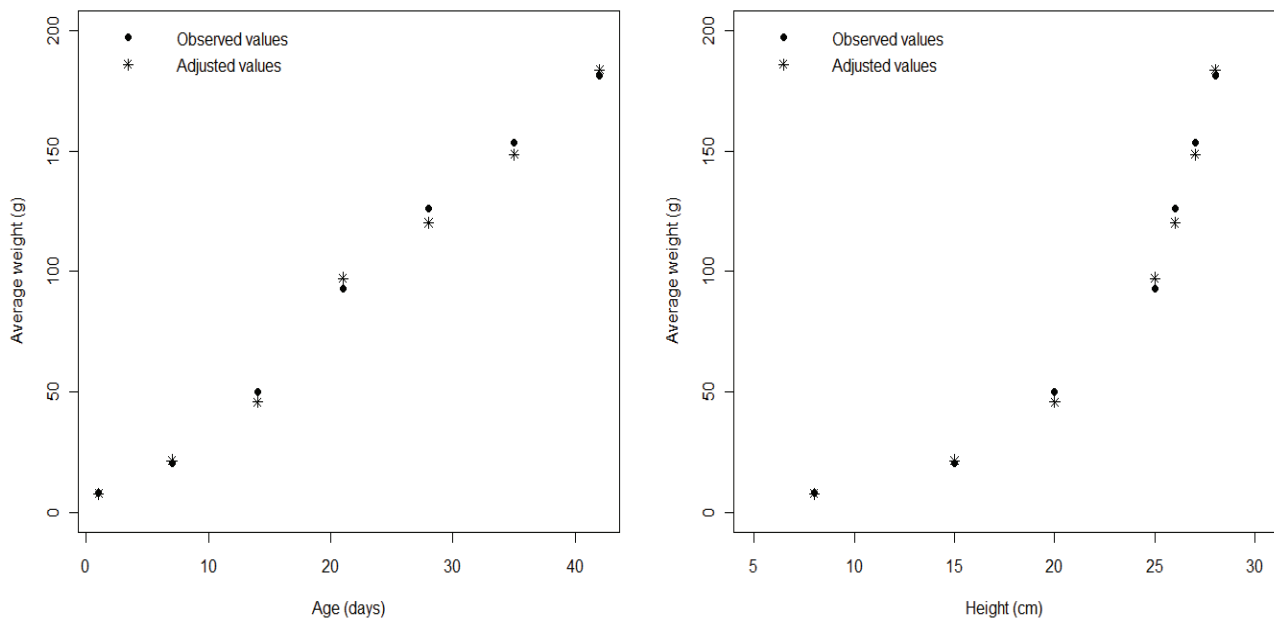


Figure 5. Fit of log-normal model.

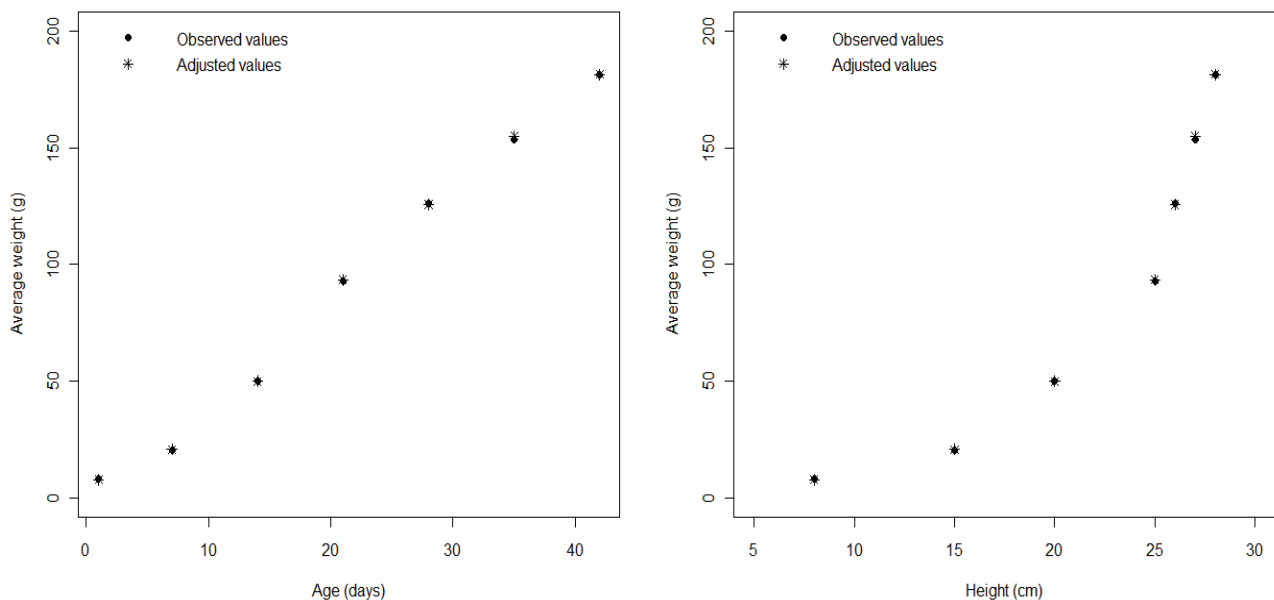


Figure 6. Fit of polynomial model.

Conclusion

The evaluation of the four goodness of fit criteria indicated that the model best fit to explain the behavior of quail's weight in relation to age and height of the birds was the polynomial regression model followed by log-normal model.

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