

Multivariate principal components analysis of the morphostructural traits of West African Dwarf sheep

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Abstract

A study was conducted to determine the interdependence among conformation traits of 178 West African Dwarf sheep and to predict body weight from their independent scores using principal components analysis. Body weight and nine morphostructural traits: height at withers, body length, heart girth, rump height, hip width, chest depth, fore canon bone length, tail length, and ear length were measured. Phenotypic correlation coefficients between body weight and body dimensions ranged from 0.78 to 0.79. Anti-image correlations of the body shape characters showed that partial correlations were low. Two principal components were extracted from the factor analysis with varimax rotation of the inter-correlated traits which accounted for 69.50% of the total variance. The first principal component (PC1) explained 57.70% and second principal component (PC2) contributed 11.70% of the variance. Both PC1 and PC2 could be considered in the selection programme to obtain animals with better conformations using minimal number of measurements.

Keywords: Body dimensions, multivariate, principal components analysis, traits, WAD sheep

Introduction

Morphological traits and body weight are important to access the potentials in West African Dwarf (WAD) sheep, these traits indicates the usefulness of the WAD sheep for commercial production purpose. Researchers have given attention to the body size of livestock animals as the object being observed. Therefore, attempt to measure these traits in a way other than conventional weighing and grading seems appropriate. Multivariate techniques have been greatly applied in animal breeding research. It has been used to access body shape and size in certain animals (Brown *et al.*, 1973; Fumio *et al.*, 1982; Hammock and Shrode, 1986; Yakubu *et al.*, 2009) in cattle, Riva *et al.* (2004) and Salako (2006) in sheep and goats. Several studies have also been carried out showing a strong correlation between some linear body sizes and production traits, that is linear body size can be used to estimate the body weight of sheep (Otoikhian *et al.*, 2008; Abdel-

Moneim, 2009) and goats (Mukherjee *et al.*, 1981, Adeyinka and Muhammad, 2006; Jimmy *et al.*, 2010).

Analysis of variance and product moment correlations are widely used to characterize phenotypic and genetic relationships among traits in a breeding programme. However, principal components analysis is a valuable refinement for analyzing data on linear body measurements and performance test traits (Dunteman, 1989; Miserani *et al.*, 2002; Posta *et al.*, 2007). Principal components according to Dunteman,(1989); Johnson and Wichern (1998) are linear combinations of all the original set of variables that are estimated in such a way that the first principal component explains the largest percentage of the total phenotypic variance. The first principal component therefore is a measure for the explanation and identification of trait groups which can allow quantitative measurements for animal conformation and enable genetic parameters for these traits to

be estimated, thereby permitting its inclusion in breeding programmes. Also, according to Morrison (1976), the principal components analysis is a multivariate methodology that can be used with success when characteristics are correlated. This analysis transforms an original group variable into another group principal component which are linear combinations of all the original variables. The main advantage is the independence of these components. From the view point of animal genetics and improvement, principal components simultaneously consider a group of attributes which may be interesting for selection purpose.

The West African Dwarf sheep is one of the breeds of sheep in Nigeria, reared majorly for meat production. However, multivariate techniques have not fully been exploited in the objective description of their body conformations. This present study was therefore designed to document changes in the morphometric traits of WAD sheep and also to explore the relationships among body dimensions using principal component analysis with a view to reducing the number of body measurements required for genetic and breeding purposes. The results so obtained could be used to predict body weight in the field and for selection purpose.

Materials and Methods

Data used for this study were collected from villages around Odeda agro-ecological zone, Abeokuta, Ogun State, in South-western Nigeria. Data were obtained from one hundred and seventy eight ($n = 178$) randomly selected WAD sheep (consisting 47 rams and 131 ewes) between April and July 2011, reared under the extensive system of management. Animals sampled have age range between 1 and 3 years. Body weight and nine morphometric traits were

measured on each animal following the standard procedure and anatomical reference described by Yakubu (2009). The traits measured included: Body length (BL), wither height (WH), heart girth (HG), chest depth (CD), rump height (RH), hip width (HW), tail length (TL), fore canon bone length (FCB), and ear length (EL). A graduated measuring steel tape was used for the height measurements, while the length and circumference measurements were done using a flexible tape. In order to avoid intra-individual variations, all the measurements were taken by the same person.

Data collected were subjected to analysis of variance (ANOVA) using a General Linear Model (GLM) procedure of SPSS 2001, mean, standard deviation and coefficient of variation of each measurement were computed. The correlation coefficients of body weight and the linear body measurements were also determined. From the correlation matrix, data for the principal component factor analysis were generated. *Barlett's test* of sphericity was used to test if the correlation matrix was an identity matrix (each variable correlated with itself) or a correlation matrix full of zero. The suitability of the dataset to principal components analysis (PCA) was further tested by *Kaiser-Meyer Olkin* (KMO) measures of sampling adequacy. This tested whether the partial correlations among the variables were small. A KMO measure of 0.60 and above is considered adequate (Eyduran *et al.*, 2010).

According to Everitt *et al.* (2001), PCA is a method for transforming the variables in a multivariate dataset, X_1, X_2, \dots, X_p , into new variables Y_1, Y_2, \dots, Y_p which are uncorrelated with each other and account for decreasing proportions of the total variance of the original variables defined as:

$$\begin{aligned}
 y_1 &= a_{11} x_1 + a_{12} x_2 + \dots + a_{1p} \\
 x_p y_2 &= a_{21} x_1 + a_{22} x_2 + \dots + a_{2p} x_p \\
 y_p &= a_{p1} x_1 + a_{p2} x_2 + \dots + a_{pp} x_p
 \end{aligned}$$

with the coefficients being chosen so that y_1, y_2, \dots, y_p account for decreasing proportions of the total variance of the original variables X_1, X_2, \dots, X_p .

During the evaluation, factors were rotated with varimax rotation of Kaiser. The aim of the varimax rotation is to maximise the sum of the variances of a_{ij}^2 quadratic weight. The stepwise variable selection of multiple regression procedure was used to obtain models for predicting body weight from body measurements (a) and from established principal components (b):

$$\begin{aligned}
 BW &= a + B_1 X_1 + \dots + B_k X_k \quad \text{--- (a)} \\
 BW &= a + B_1 PC_1 + \dots + B_k PC_k \quad \text{--- (b)}
 \end{aligned}$$

where BW is the body weight, 'a' is the intercept; B_i is the i -th partial regression coefficient of the i -th linear body measurements, X_i or the i -th principal component. The factor programme of SPSS (2001) was used for the analysis.

Results

The descriptive statistics (means, standard deviations and coefficients of variation of body weight and conformation traits) of WAD sheep are presented in Table 1. It shows the changes in the morphometric traits of the WAD sheep from age one to age three years.

Table 2 presents the anti-image matrices (anti-image correlation) of the linear body measurements of the WAD sheep. Negative relationships existed among the body

measurements except for chest depth and fore canon bone length (0.158), hip width and fore canon bone length (0.092), ear length and hip width (0.295), ear length and wither height (0.115), heart girth and wither height (0.077). The highest correlation was between ear length and hip width ($r = 0.295$). The anti-image correlations computed showed that partial correlations were low. Kaiser-Meyer-Olkin measures of sampling adequacy computed indicated that the sample sizes were adequate to apply PCA (Kaiser, 1960). Kaiser-Meyer-Olkin measures of sampling adequacy computed for the WAD sheep was 0.886. The overall significance of the correlation matrices tested with Barlett's Test of sphericity for the body dimensions of the WAD sheep ($\chi^2 = 619.296$; $P < 0.01$) provided enough support for the validity of the factor analysis of the dataset. The communalities which stand for the proportion of the variance in the original variables is accounted for by the factor solution and it ranged from 0.761 to 0.767.

Table 3 shows the total variance, here, principal component 1 (PC1) accounts for 57% of the variance and principal component 2 (PC2) accounts for 11% and so on. As expected, the sum of the Eigen value is equal to the number of variables. The Eigen values showed the amount of variance out of the total variance explained by each of the factors. The first factor (5.196) accounts for the most variance and hence has the highest Eigen value, and the next factor (1.056) accounts for much of the leftover variance. The sixth column (% of variance) has the percentage of total variance accounted for by each factor (i.e. factor 1 accounts for 57.734 and factor 2 accounts for 11.736. The second row shows a value of 69.470 which is the sum of 57.734 and 11.736. Table 4 shows the

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Table 1. Descriptive statistics of body measurements of WAD sheep (age 1-3years)

Traits	Age	Mean	±	Standard deviation	Coefficient of variation	of N
Body weight (kg)	1	47.49	±	3.18	6.43	26
	2	52.99	±	5.95	11.23	47
	3	56.08	±	3.82	6.81	66
Body length (cm)	1	11.71	±	1.43	12.21	26
	2	54.79	±	4.22	7.70	47
	3	58.32	±	3.58	6.14	66
Wither height (cm)	1	23.35	±	2.63	11.26	26
	2	13.67	±	1.64	11.99	47
	3	14.22	±	1.57	11.04	66
Rump height (cm)	1	10.86	±	1.63	15.00	26
	2	26.34	±	2.97	11.27	47
	3	29.23	±	3.41	11.66	66
Fcb length (cm)	1	9.14	±	1.47	16.08	26
	2	11.96	±	2.28	19.06	47
	3	13.87	±	1.89	13.63	66
Chest depth (cm)	1	52.81	±	4.83	9.15	26
	2	10.18	±	1.35	13.26	47
	3	10.77	±	1.71	15.87	66
Hip width (cm)	1	19.19	±	2.29	11.93	26
	2	59.85	±	6.97	11.65	47
	3	68.00	±	4.87	7.16	66
Ear length (cm)	1	4.29	±	0.89	20.75	26
	2	4.50	±	1.35	30.00	47
	3	4.10	±	0.94	22.93	66
Heart girth (cm)	1	62.02	±	5.64	9.09	26
	2	65.72	±	7.72	11.75	47
	3	73.52	±	7.44	10.12	66

Table 2. Anti-image matrices of WAD sheep

Parameter	Body weight	Body length	Wither height	Rump height	Fcb length	Chest length	Hip width	Ear length	Heart girth
Body weight (kg)	0.901 ^a								
Body length (cm)	-0.176	-0.891 ^a							
Wither height (cm)	-0.081	-0.183	0.885 ^a						
Rump height (cm)	-0.002	-0.402	-0.302	0.903 ^a					
Fcb length (cm)	-0.026	-0.198	-0.239	-0.135	-0.760 ^a				
Chest depth (cm)	-0.106	-0.140	-0.315	-0.129	0.158	0.904 ^a			
Hip width (cm)	-0.193	-0.082	0.018	-0.162	0.092	-0.119	0.893 ^a		
Ear length (cm)	-0.257	-0.138	0.115	-0.081	-0.219	-0.084	0.295	0.839 ^a	
Heart Girth (cm)	-0.424	-0.180	0.077	-0.068	-0.232	-0.178	-	-0.084	0.892 ^a
							0.234		

component matrix made up of PC1 and PC2. Each number represents the quantification of the body measurements and the unrotated factor. For example, the quantification of body weight and PC1 was 0.872, while that of heart girth and PC1 was 0.876 which was the highest quantification value observed and this implies that PC1 has a high significant quantification with majority of the traits observed body weight (0.872), body length (0.840), wither height (0.781), rump height (0.869), chest depth

(0.742), hip width (0.707) and heart girth (0.876); implying that there is a high quantification in the variables.

Table 5 provides the model summary which gives the simple correlation (R) and R² value which describes how much of the dependent variable (body weight) can be explained by the independent variable (heart girth). Table 6 described the coefficients of the variables which gives information on each predictor variable. This provides the information necessary to

Table 3. Total variance of each component showing the percentage of variance and cumulative percentage

Component	Initial Eigen values			Extraction sums of squared loadings			Rotation		
	Total	% of variance	Cumulative %	Total	% of variance	Cumulative %	Total	% of variance	Cumulative %
1	5.196	57.734	57.734	5.196	57.734	57.734	4.394	48.823	48.823
2	1.056	11.736	69.470	1.056	11.736	69.470	1.858	20.647	69.470
3	0.703	7.815	77.285	0.703	7.815	77.285	0.703	7.851	77.285
4	0.574	6.382	83.667	0.574	6.382	83.667	0.574	6.382	83.667
5	0.509	5.656	89.323	0.509	5.656	89.323	0.509	5.656	89.323
6	0.307	3.410	92.733	0.307	3.410	92.733	0.307	3.410	92.733
7	0.271	3.006	95.739	0.271	3.006	95.739	0.271	3.006	95.739
8	0.200	2.227	97.966	0.200	2.227	97.966	0.200	2.227	97.966
9	0.183	2.034	100.000	0.183	2.034	100.000	0.183	2.034	100.000

Table 4. Component matrix

Parameter	Component	
	1	2
Body weight (kg)	0.872	-0.027
Body length (cm)	0.840	-0.106
Wither height (cm)	0.781	-0.022
Rump height (cm)	0.869	-0.043
Fore canon bone length (cm)	0.455	0.706
Chest depth (cm)	0.742	-0.201
Hip width (cm)	0.707	-0.452
Ear length (cm)	0.585	0.546
Heart girth (cm)	0.876	-0.006

predict body weight from heart girth and other body measurements and they also contribute significantly to the models.

Discussion

The positive and significant correlation among the body measurements indicates a high predictability among the variables (Pundir *et al.*, 2011). The anti-image correlations computed showed that partial correlations were low, indicating that true factors existed in the data. This is buttressed by Kaiser-Meyer Olkin measure of sampling adequacy studied from the diagonal of partial correlation, revealing the proportion of the variance in the body

measurements caused by the underlying factor of 0.886. The overall significance of the correlation matrices tested with *Barlett's Test* of sphericity for the body dimensions of the WAD sheep ($\chi^2 = 619.269, P < 0.01$) provided enough support for the validity of the factor analysis of the dataset. After the varimax rotation, most of the variables that measured the linear body dimensions (body length, wither height, rump height, chest depth, hip width and heart girth) were highly significantly correlated. This is because the factors were extracted successfully and they accounted for gradual reduction in the overall variance. These findings are in consonance with the observations of Shahin and Hassan (2001) in their studies with New Zealand White rabbits. Similarly, Shahin *et al.* (1993) reported for Egyptian buffalo that the first factor was characterised by high positive loadings on all body shape characters, the variables associated with body length had the highest loading followed by heart girth and abdominal girth. It also corroborates the earlier submission of Salako (2006) that the first principal component comprising seven measurements (foreleg length, tail length, face length, rump height, withers height, body length and heart girth) explained 67.7% of the generalised variance and can be considered as a generalised size factor. Similarly, results of this study were consistent with those reported by McCracken *et al.* (2000) in ducks, and in

Table 5. Model summary*

Model	R	R ²	Adjusted R ²	Standard error of the estimate
1	0.803	0.645	0.642	3.863
2	0.828	0.686	0.681	3.646
3	0.837	0.701	0.693	3.577
4	0.847	0.717	0.706	3.496

* 1 = Predictor: (constant), heart girth, 2 = Predictors: (constant), heart girth, body length, 3 = Predictors: (constant), heart girth, body length, ear length, 4 = Predictors: (constant), heart girth, body length, ear length, hip width and R = Regression value.

Table 6: Coefficients of the variables

Model		Unstandardized Coefficients		Standardized Coefficients		
		B	Std. Error	Beta	t	Sig.
1	(Constant)	-24.712	2.727		-9.060	.000
	heart girth	.728	.051	.803	14.256	.000
2	(Constant)	-26.586	2.621		-10.144	.000
	heart girth	.549	.067	.605	8.158	.000
	body length.	.200	.052	.284	3.832	.000
3	(Constant)	-28.691	2.728		-10.519	.000
	heart girth	.506	.068	.558	7.388	.000
	body length.	.178	.052	.253	3.416	.001
	ear length	.623	.270	.141	2.311	.023
4	(Constant)	-29.055	2.670		-10.884	.000
	heart girth	.422	.075	.465	5.619	.000
	body length.	.142	.053	.202	2.681	.008
	ear length	.801	.273	.181	2.934	.004
	hip width	.510	.205	.173	2.487	.014

Dependent variable: body weight

1st Model: Body weight = -24.712 + 0.728 heart girth

2nd Model: Body weight = -26.586 + 0.549 heart girth + 0.200 body length

3rd Model: Body weight = -28.691 + 0.506 heart girth + 0.178 body length + 0.623 ear length

4th Model: Body weight = -29.055 + 0.422 heart girth + 0.142 body length + 0.801 ear length + 0.510 hip width

ruminants by (Yakubu, 2009; Yakubu *et al.*, 2009) where PC1 was termed the overall body size. Similarly, Kashiwamura *et al.* (2001) reported that the first factor accounted for the largest variance in horses. The principal component could be important in evaluating animals for breeding and selection purposes. Both principal components extracted could play a role in the ranking of the animals and thus provide an opportunity to select the animals based on a group of variables rather than on

isolated traits. Gusmao-Filho *et al.* (2009), extracted five principal components from eleven original traits and concluded that these could be of great importance in the determination of body attributes of sheep for both reproduction and meat production.

The interdependent original body measurements or conformation traits and their independent principal component scores were used to predict body weight of sheep. The results of the stepwise multiple regression analysis revealed that heart girth

was the single variable of utmost importance in the prediction of body weight. Heart girth alone explained about 64% of the variation in body weight. When body length was added to the model, the proportion of explained variance increased to 68%. The combination of heart girth, body length and ear length accounted for about 70%. This result indicates that body weight can be predicted with a fair degree of accuracy from body dimensions. Similar findings of this nature have been reported by (Yakubu, 2009; Yakubu *et al.*, 2009). However, the use of body measurements to predict body weight or growth should be treated with caution due to multicollinearity which has been shown to be associated with unstable regression estimates (Ibe, 1989), thereby leading to unreliable predictions. This justifies the use of indices of the body measurements referred to as principal components for prediction, since they are orthogonal to each other.

Conclusion

In WAD sheep and according to the results of this study, PC1 had the largest share of the total variance and correlated highly with body length, wither height, rump height, heart girth, chest depth and hip width. These highly correlated traits and the use of orthogonal body shape characters (PC1 and PC2) could be used to predict the body weight of the WAD sheep, which could be more reliable when compared to the original inter-correlated body measurements. Multiple collinearity of interdependent explanatory variables could result in erroneous conclusions when original body measurements are used as predictors. The PC2 was orthogonal to PC1 and loaded heavily on the fore canon bone length. The multivariate principal component analysis technique was also

used to consolidate and describe the interdependence (general size and body indices) of WAD sheep. This method of analysis can therefore be applied in the ranking of animals based on different indices thus aiding in the drastic reduction of the number of body measurements required for selection in a breeding programme for improvement and performance of the WAD sheep.

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