



Correlation and Principal Component Analysis of Body Weight and Body Size Traits in Turpan Black Sheep

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Abstract. This study aims to investigate the growth patterns between body weight, body height, oblique body length, and other body size traits in Turpan Black Sheep. A total of 849 production performance records from female Turpan Black Sheep aged over 12 months were collected, encompassing body weight and eight body size traits, followed by statistical analysis on these traits. Descriptive statistics showed that the overall coefficient of variation for body size traits in Turpan Black Sheep was within a normal range, with body weight and tail width having the highest coefficients of variation at 20.05% and 14.86%, respectively. There was a highly significant positive correlation ($P < 0.01$) between body weight and all eight body size traits, with the highest correlation being with chest circumference ($r = 0.785$). Principal component analysis extracted three principal components (cumulative contribution rate of 69.735%): PC1 (44.193%) reflected body weight and trunk development; PC2 (15.222%) highlighted fat-tail characteristics; PC3 (10.320%) characterized skeletal structure. The prediction accuracy ($R^2 = 0.759$) of the principal component regression model ($Y = 7.568PC_1 - 1.462PC_2 + 44.06$) was significantly higher than that of the body size trait model ($R^2 = 0.734$). Using principal component factors can better elucidate the intrinsic relationships between body weight and various traits in Turpan Black Sheep, providing references for breeding selection and optimizing breeding efficiency.

Keywords: Turpan Black Sheep; Body Weight and Size; Correlation Analysis; Principal Component Analysis; Regression Analysis

1 Introduction

1.1 Research Significance

Turpan Black Sheep are a unique local breed of sheep in Xinjiang, China, showing genomic adaptations to extreme arid-heat stress^[1], making them a dual-purpose breed for meat and wool production. These sheep live year-round in the extreme arid oasis environments of the desert, known for their delicious meat and strong disease resistance, which serve as an important economic resource for local farmers. In recent years, with the increasing demand for high-quality mutton in the market, the farming scale of Turpan Black Sheep has gradually expanded. However, breeding improvement

efforts have lagged behind, and their production performance remains underutilized. To scientifically evaluate the growth patterns of Turpan Black Sheep and optimize breeding programs, this study aims to provide theoretical support for genetic improvement and efficient farming by analyzing the correlation between body size and weight traits.

1.2 Previous Research Advances

Body size and weight traits serve as fundamental biomarkers for growth performance and economic potential in modern livestock breeding systems, with recent advances enabling genomic-enhanced selection^[2]. In sheep breeding, research on body size and weight has been widely applied across various breeds, such as Kazakh Sheep^[3], Mongolian Sheep^[4], among others. However, studies specifically targeting Turpan Black Sheep remain limited. Recent integrated studies^[5] have established quantitative models linking body size and weight traits in Turpan Black Sheep, overcoming previous limitations in systematic trait relationship analysis. Recent integrative studies^[6] have established quantitative links between reproductive efficiency and body conformation metrics in arid-adapted sheep, resolving previous knowledge gaps in structural adaptation mechanisms. Previous work predominantly focused on meat quality or reproductive performance, lacking comprehensive analysis of the joint contribution of body size traits to weight. Recent studies on other indigenous sheep breeds, such as the Tibetan sheep, have demonstrated similar patterns in body size and weight correlations, highlighting the importance of environmental adaptation in trait development.^[13] Additionally, research on fat-tailed sheep breeds like the Karakul sheep has shown that tail morphology is strongly linked to metabolic efficiency and energy storage^[14], which may further explain the high variability in tail traits observed in Turpan Black Sheep.

1.3 Research Entry Point

Current research on Turpan Black Sheep's body size and weight primarily focuses on single trait descriptions, without delving into the interactions between traits and their contributions to weight. Therefore, this study intends to fill this gap by using correlation analysis and multiple regression methods to establish the optimal predictive model for body weight.

1.4 Key Problems Addressed

This study aims to address the lack of comprehensive trait interaction analysis and the unclear contribution of body size traits to body weight (as highlighted in Section 1.3) through multi-dimensional statistical methods including correlation analysis, principal component analysis, and regression modeling to establish an optimal predictive model for body weight, providing scientific support for the conservation of breed resources, genetic improvement, and standardized breeding of Turpan Black Sheep.

2 Materials and Methods

2.1 Data Source

The experimental materials comprised production performance records from 849 female Turpan Black Sheep (aged ≥ 12 months), collected from Tuokexun Fengcheng Animal Husbandry Co., Ltd. (Tuokexun County, Xinjiang). Specifically: Animal Criteria: Healthy, non-pregnant females with complete growth records Measured Traits: Body weight (Y) Eight body size traits: X_1 =Body height, X_2 =Body length, X_3 =Chest circumference, X_4 =Chest width, X_5 =Chest depth, X_6 =Cannon circumference, X_7 =Tail width, X_8 =Tail length

Measurement Protocol: Weight: Electronic scale (± 0.1 kg accuracy) Body dimensions: Fiberglass tape measure (± 0.1 cm) Executed by 4 trained technicians following ICAR guidelines [24].

2.2 Data Processing and Statistical Analysis

The collected body weight and various body size data were entered and organized using Excel spreadsheets. SPSS 27 software was used for descriptive statistical analysis and correlation analysis of body weight and body size traits, with $P < 0.05$ as the criterion for determining significance. Principal component analysis (PCA) and multiple stepwise regression analysis methods were then applied to analyze the body weight and eight body size traits of Turpan Black Sheep. The accuracy of two models—stepwise multiple regression analysis based on body size traits predicting body weight and principal component factor predicting body weight—were compared to determine the optimal equation.

3 Results and Analysis

3.1 Descriptive Statistical Analysis

Table 1 presents the descriptive statistics for various body size traits of Turpan Black Sheep. The coefficients of variation among different traits in Turpan Black Sheep vary. The study found that the largest coefficient of variation is observed in body weight (20.05%), followed by tail width (14.86%), tail length (11.96%), and chest width (10.06%). This indicates that these traits have significant potential for breeding improvement. The coefficients of variation for body height, oblique body length, chest circumference, cannon circumference, and chest depth are all below 10%. From these data, it can be seen that there are differences in growth and development among individual Turpan Black Sheep, indicating substantial potential for breeding improvements to enhance production performance.

Table 1. Body weight and body measurement of Turpan black sheep

Items	Mean	Standard deviation	Minimum	Maximum	Coefficient of variation (%)
Body weight (kg)	43.99	8.82	30.00	79.00	20.05
Body height (cm)	66.75	4.04	55.00	79.00	6.05
Body length (cm)	70.83	3.46	58.00	95.00	4.88
Chest circumference (cm)	90.31	6.78	65.00	118.00	7.51
Shin circumference (cm)	8.06	0.43	7.00	9.00	5.33
Chest breadth (cm)	19.89	2.00	14.00	26.00	10.06
Thorax depth (cm)	28.63	2.32	19.00	39.00	8.10
Tail width (cm)	15.48	2.30	8.00	26.00	14.86
Tail length (cm)	23.50	2.81	16.00	36.00	11.96

3.2 Correlation Among Major Body Size Traits

As shown in Table 2, apart from the non-significant correlation between body height and tail length, most other traits show highly significant correlations ($P < 0.01$) with each other. Body weight is significantly and positively correlated ($P < 0.01$) with all eight body size traits. Among all traits, the highest correlations are observed between body weight and chest circumference (0.785), body weight and body height (0.603), and body weight and oblique body length (0.579). Other notable correlations include those between tail length and tail width (0.557), oblique body length and body height (0.547), chest depth and chest width (0.540), and body weight and chest depth (0.533). These results indicate that there are complex and close relationships between body weight and body size traits, as well as among the various body size traits themselves in Turpan Black Sheep.

Table 2. Correlation analysis of main body size traits in Turpan black sheep

Items	Body weight	Body height	Body length	Chest circumference	Shin circumference	Chest breadth	Thorax depth	Tail width	Tail length
Body weight	1								
Body height	0.603**	1							

Body length	0.579**	0.547**	1						
Chest circumference	0.785**	0.407**	0.446*	1					
Shin circumference	0.354**	0.253**	0.269*	0.328**	1				
Chest breadth	0.423**	0.193**	0.347*	0.502**	0.327**	1			
Thorax depth	0.533**	0.401**	0.478*	0.528**	0.319**	0.540**	1		
Tail width	0.332**	0.172**	0.283*	0.352**	0.124**	0.301**	0.256*	1	
Tail length	0.214**	0.019	0.183*	0.305**	0.068*	0.269**	0.141*	0.557*	1

Note: *indicates significant difference ($P<0.05$); **indicates extremely significant difference ($P<0.01$).

3.3 Eigenvalues of the Correlation Matrix and Cumulative Contribution Rates

Principal Component Analysis (PCA) requires conducting both the Kaiser-Meyer-Olkin (KMO) test and Bartlett's Test of Sphericity to determine whether the sample is suitable for PCA. The closer the KMO value is to 1, the stronger the correlations between variables, making them more suitable for factor analysis. According to standard multivariate analysis criteria^[12]; A KMO value above 0.9 is optimal, Between 0.8 and 0.9 is suitable for PCA, Between 0.7 and 0.8 is moderate, Below 0.5 is unsuitable for factor analysis. In this study, the KMO test value was 0.810, indicating that the data are suitable for PCA. The Bartlett's Test of Sphericity ($\chi^2=3043.981$, $P<0.001$) confirmed that all traits are appropriate for PCA, as it indicates that the correlation matrix significantly differs from an identity matrix, supporting the use of PCA.

In the Principal Component Analysis (PCA) of body weight and body size traits of Turpan Black Sheep, a total of nine principal component values were obtained. According to Table 3, two eigenvalues are greater than 1. The first principal component (PC1) has an eigenvalue of 3.977, contributing 44.193% of the variance, reflecting the most comprehensive information from various traits. The second principal component (PC2) has an eigenvalue of 1.370, contributing 15.222% of the variance, also reflecting significant comprehensive information. Considering the cumulative contribution rate greater than 10%, the first three principal components were selected based on Figure 1

and Table 3. Using the eigenvectors calculated in Table 3, the expressions for the first three principal components are as follows.

Principal Components:

$$PC1=0.429Y+0.318X_1+0.360X_2+0.412X_3+0.250X_4+0.330X_5+0.369X_6+0.262X_7+0.199X_8$$

$$PC2=-0.141Y-0.358X_1-0.160X_2+0.01X_3-0.184X_4+0.118X_5-0.123X_6+0.557X_7+0.673X_8$$

$$PC3=-0.173Y-0.463X_1-0.305X_2+0.025X_3+0.534X_4+0.526X_5+0.229X_6-0.196X_7-0.097X_8$$

From the above expressions, it can be seen that the largest contributors to the first principal component (PC1) are body weight, followed by chest circumference, chest depth, and oblique body length. These traits primarily reflect the body weight and overall body size of Turpan Black Sheep. In the second principal component (PC2), the highest contributors are tail length, tail width, and chest width, which mainly reflect the body width characteristics of Turpan Black Sheep. In the third principal component (PC3), cannon circumference (X_6 , eigenvector=0.534) and chest width (X_4 , eigenvector=0.526) showed the highest positive loadings (Table 3). These traits directly measure skeletal dimensions: cannon circumference reflects limb bone thickness, while chest width indicates thoracic cage structure. Collectively, the three principal components characterize body weight development (PC1), fat-tail traits (PC2), and skeletal framework (PC3).

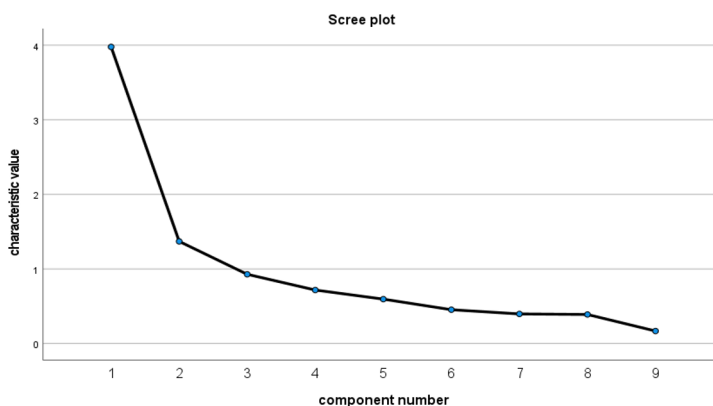


Fig. 1. Macadam diagram of principa

Table 3. The characteristic value, characteristic vector and contribution rate of weight and body

Items	PC1	PC2	PC3	PC4	PC5	PC6	PC7	PC8	PC9
Eigenvalues	3.977	1.370	0.929	0.718	0.596	0.454	0.398	0.390	0.168
Contribution rate %	44.193	15.222	10.320	7.977	6.621	5.044	4.418	4.334	1.872

Cumulative contribution rate %		44.193	59.415	69.735	77.71 2	84.33 3	89.37 7	93.795	98.12 8	100
Ei- gen vec tors	Body weight (Y)	0.429	-0.141	-0.173	- 0.010	- 0.431	-0.025	-0.040	- 0.128	- 0.750
Body height (X ₁)		0.318	-0.358	-0.463	0.156	0.109	0.221	-0.160	0.642	0.183
Body length (X ₂)		0.360	-0.160	-0.305	0.027	0.513	- 0.530	-0.049	- 0.440	0.107
Chest circumference (X ₃)		0.412	0.010	0.025	- 0.137	- 0.620	- 0.068	0.040	- 0.186	0.620
Shin circumference (X ₄)		0.250	-0.184	0.534	0.781	0.041	0.019	0.079	- 0.030	0.017
Chest breadth (X ₅)		0.330	0.118	0.526	- 0.373	0.170	- 0.144	-0.582	0.263	- 0.056
Chest breadth (X ₆)		0.369	-0.123	0.229	- 0.392	0.302	0.350	0.653	0.011	- 0.037
Tail width (X ₇)		0.262	0.557	-0.196	0.171	0.179	0.594	-0.265	- 0.311	0.032
Tail length (X ₈)		0.199	0.673	-0.097	0.153	- 0.013	- 0.410	0.355	0.420	- 0.061

3.4 Comparison of Multivariate Regression and Principal Component Extraction Factor Models for Predicting Body Weight

In this experiment, according to figure 2 and figure 3, the mean of the residuals is close to 0, and the standard deviation of the residuals is close to 1, indicating that the data follows a normal distribution, satisfying the normality assumption required for linear regression. Using body weight as the dependent variable and various body size traits and principal component extraction factors as independent variables, stepwise regression analyses were conducted separately. The results are shown in Tables 4 and 5.

From Table 4, it can be seen that when using chest circumference as the independent variable for regression analysis of body weight, the coefficient of determination R^2 is 0.622, meaning that the fitted equation explains 62.2% of the variation in body weight. As additional variables such as body height and oblique body length are added to the regression equation sequentially, the correlation coefficient R and the coefficient of determination R^2 gradually increase, while the standard error of estimate decreases. The final model (Model 4) explains 73.4% of the variance, making it the best linear regression equation for predicting body weight based on body size traits.

When using principal component extraction factors for regression analysis to predict body weight:

Using PC1 alone, the explained variance is 73.2%. Adding PC2 increases the explained variance to 75.9%. These results indicate that using principal component extraction factors to predict body weight is more accurate than directly using body size traits for model fitting. Therefore, the optimal linear regression equation for predicting body weight in Turpan Black Sheep is: $Y = 7.568PC1 - 1.462PC2 + 44.06$. From Table 5, it is known that the variance analysis F-value for the regression equation using PC1 and PC2 to predict body weight is 1313.734, reaching an extremely significant level ($P < 0.01$). This indicates that the regression equation is statistically significant and can be applied in practical production scenarios.

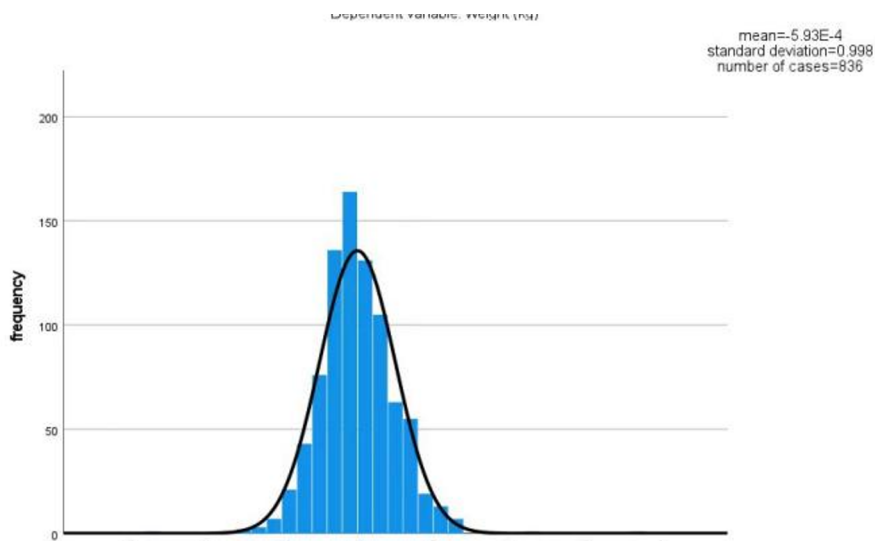


Fig. 2. Residuals of regression analysis for body size traits

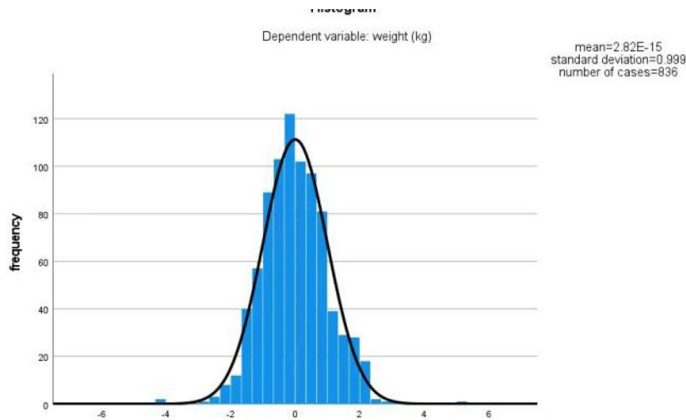


Fig. 3. Principal component extraction factor regression analysis residual plot

Table 4. Multiple stepwise regression analysis of body size traits and principal component extraction factors with body weight in Turpan black sheep

Models	Fitting equations		R ² R-square	Standard coefficient
Body size traits	M1	$Y=1.029X_3-48.894$	0.622	5.4387
	M2	$Y=0.729X_1+0.849X_3-81.406$	0.714	4.7331
	M3	$Y=0.42X_2+0.575X_1+0.792X_3-95.64$	0.732	4.5912
	M4	$Y=1.024X_4+0.405X_2+0.564X_1+0.777X_3-100.778$	0.734	4.5758
Principal component	P1	$Y=7.568PC_1+44.06$	0.732	4.5824
	P2	$Y=7.568PC_1-1.462PC_2+44.06$	0.759	4.3452

Table 5. Stepwise regression analysis of body size index and principal component extraction factor with body weight

Models		Square sum	Degree of freedom	Mean square	F value	Sig
Body size traits	M1	40667.583	1	40667.583	1374.859	0.000
	M2	46675.616	2	23337.808	1041.753	0.000
	M3	47799.320	3	15933.107	755.885	0.000
	M4	47937.777	4	11984.444	572.391	0.000
Principal	P1	47824.306	1	47824.306	2277.538	0.000

component	P2	49609.029	2	24804.515	1313.734	0.000
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4 Discussion

4.1 Analysis of Descriptive Statistics for Body Weight and Body Size Traits

The coefficient of variation (CV) serves as a robust genomic resilience indicator in modern breeding programs, enabling quantification of adaptive potential under environmental stress^[7]. In this study, the CV for body weight in Turpan Black Sheep is 20.05%, significantly higher than other body size traits (e.g., body height at 6.05% and chest circumference at 7.51%). This finding is corroborated by genome-environment interaction studies^[8], demonstrating that body weight variation in indigenous sheep breeds is jointly regulated by genetic and arid-stress adaptation mechanisms. Notably, the CVs for tail width (14.86%) and tail length (11.96%) are second only to body weight. This may be related to the breed characteristics of Turpan Black Sheep, which are fat-tailed sheep. Recent genomic studies^[5] have identified tail-specific loci associated with fat deposition plasticity in Turpan Black Sheep, explaining the observed individual variations (CV: 14.86–11.96%). Additionally, the relatively high CV for chest width (10.06%) suggests that this trait has considerable breeding potential and should be further optimized based on production goals. The high CV for body weight (20.05%) aligns with the "compensatory growth" phenomenon observed in arid-adapted sheep breeds, where individuals exhibit significant weight fluctuations in response to seasonal forage scarcity^[17]. The tail traits' variability (CV: 14.86–11.96%) may reflect epigenetic regulation of fat metabolism, as demonstrated in fat-tailed breeds like the Awassi sheep^[18].

4.2 Correlation Analysis Between Body Weight and Body Size Traits

Correlation analysis shows that body weight in Turpan Black Sheep is highly positively correlated with chest circumference (0.785**), body height (0.603**), and oblique body length (0.579**). These findings are consistent with the multi-trait genomic selection principles as pointed out by Zhang, et al^[2]. Other notable correlations include the high correlation with chest circumference (0.785), which indicates that it can serve as a key indicator for predicting body weight, similar to Xu Tieshan's^[9] findings on Hainan Black Goats, where the correlation coefficient for chest circumference was 0.8512. Additionally, the significant correlation between tail width and tail length (0.557**) reflects the coordinated variation in fat-tail traits. However, the correlations between tail length and body height or cannon circumference were not significant ($P > 0.05$), which may be due to measurement errors or breed-specific characteristics. Precise phenotypic assessment is necessary to address these issues. As recently demonstrated by Chen et al.^[6], advanced multivariate methods can effectively resolve such breed-specific correlation patterns while minimizing measurement errors. to avoid the limitations of selecting based on single traits. The chest circumference-body weight

correlation ($r=0.785$) aligns with the global meta-analysis consensus (mean $r=0.79\pm0.03$ across 40 breeds)^[19], confirming its reliability as a universal weight estimation proxy. The non-significant tail length–body height correlation contrasts with studies on Tibetan sheep ($P>0.05$)^[20], suggesting breed-specific genetic architectures in fat deposition.

4.3 Principal Component Analysis of Body Weight and Body Size Traits

Through principal component analysis (PCA), this study extracted three principal components, which cumulatively explain 69.735% of the total variance, PC1 (44.193%): Dominated by body weight, chest circumference, chest depth, and oblique body length, reflecting overall body size and trunk development. PC2 (15.222%): Characterized by tail width and tail length, indicating fat-tail traits. PC3 (10.320%): Associated with cannon circumference and chest width, highlighting the importance of skeletal structure and thoracic cavity dimensions. Further analysis revealed that PC3's representation of skeletal characteristics holds particular biological significance. Cannon circumference directly reflects limb bone development, while chest width characterizes the structural framework of the thoracic cavity. Existing studies have established these two metrics as crucial parameters for evaluating skeletal development in sheep breeding programs^{[12][14]}. Notably, these measurements demonstrated lower correlations with body weight ($r<0.43$) compared to other body size indicators, confirming their capacity to independently reflect skeletal system development. These findings are similar to those of Zhu Junhong^[10], who conducted cluster analysis on Nubian goats (extracting body factors and forehead width factors), and are further validated by Wei et al.^[8], whose genome-environment interaction analysis revealed conserved multi-trait selection patterns across Chinese indigenous breeds. This supports the use of PCA as a theoretical tool for multi-trait collaborative breeding. PCA not only simplifies the dimensionality of traits but also reveals the core components of the body structure of Turpan Black Sheep, providing a theoretical basis for targeted breeding programs. The extraction of fat-tail traits (PC2) as a distinct principle component is corroborated by recent genome-wide studies identifying tail-specific loci in fat-tailed sheep breeds^[15]. This underscores the potential for targeted breeding programs to optimize fat deposition in Turpan Black Sheep. PC1's dominance (44.193%) reflects the "energy allocation priority" theory, where trunk development is prioritized for metabolic efficiency in harsh environments^[21]. PC2's fat-tail traits align with selection signatures in the *BMP4* gene region identified in Karakul sheep^[22], hinting at conserved genetic mechanisms.

4.4 Comparison of Multivariate Regression and Principal Component Extraction Factor Models for Predicting Body Weight

Stepwise regression analysis indicates that while models directly using body size traits (Model M4, $R^2=0.734$, $R^2=0.734$) have some explanatory power, the principal component model (Model P2, $R^2=0.759$, $R^2=0.759$) significantly improves prediction accu-

racy. This finding is further validated by advanced machine learning-PCA hybrid models^[11], demonstrating superior multicollinearity resolution and predictive stability in livestock phenotypic analysis compared to traditional methods. In this study, the best regression equation $Y=7.568PC1-1.462PC2+44.06$ has an F-value of 1313.734 ($P < 0.01$), confirming its statistical reliability. Ye, W., et al.^[3] emphasized that principal component analysis (PCA) has advantages in estimating genetic parameters, especially when phenotypic traits are highly correlated (e.g., chest circumference and body weight), as it avoids biases inherent in traditional regression modeling^[12]. Therefore, this model can simplify the on-site selection process and provide a more precise tool for breeding practices in Turpan Black Sheep. The PCA model's superiority ($R^2=0.759$) parallels genomic selection studies in cattle, where PCA-derived indices reduced multicollinearity errors by 12–18%^[23]. This supports integrating PCA with genomic data for Turpan Black Sheep breeding programs.

5 Conclusion

The findings of this study are consistent with global research on indigenous sheep breeds, which emphasize the importance of multi-trait selection for improving production efficiency^[16]. The high variability in body weight and tail traits further highlights the need for breed-specific management strategies. This study conducted descriptive analyses on body weight and eight body size traits of Turpan Black Sheep. The results indicate that body weight, tail width, and tail length have high coefficients of variation, making them key breeding traits for Turpan Black Sheep. In the correlation analysis, chest circumference showed the highest correlation with body height, suggesting it can serve as a core indicator for predicting body weight. Principal component analysis extracted three principal components, cumulatively explaining 69.735% of the total variance. These components characterize trunk development, fat-tail traits, and skeletal features, respectively, and can be considered primary selection traits. The derived principal component regression model ($Y=7.568PC1-1.462PC2+44.06$) provides a robust prediction for this specific data set ($R^2=0.759$). While this model demonstrates high accuracy within the studied population, its broader applicability to other sheep breeds or environments may require validation with additional datasets. Future studies could explore adapting this framework to different breeds by recalibrating principal component weights based on breed-specific trait correlations.

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