# Exploring Body Dimensions to Predict Scale Weight of Teenagers in Imo State, South-East, Nigeria.

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Abstract- In this work, the multiple linear regression models were used to explore the relationship in body dimensions and to predict scale weight for teenagers in Imo state, Nigeria. The scale weight was predicted via anthropometric measurements as well as age, height and gender for 604 physically active secondary school students-283 boys and 321 girls within the age range of 10-21 years (in completed years). The work resulted in twelve different linear regression models. The ordinary regression procedure was used to obtain the first nine models while the forward stepwise selection procedure was employed to check the performance of first three models. Studying the effects of skeletal diameter, girth circumference measurement and gender, on the response variable (scale weight). The dataset was split accordingly and the results of the individual models were compared with the full model. The goodness and adequacy of the fitted models were studied on the basis of F-value, t-value, R-squared criterion, standard error of estimate and Mallow's C-P statistics. Results of the analysis showed that body scale weight is better estimated by anthropometric, skeletal and girth variables as well as age, height and gender than any of the reduced sets of variables. The result also indicated that all the models were of good fit and adequate.

Index Terms: Anthropometric Data, Body scale weight, forward stepwise selection procedure, Model Comparison, Multiple Linear Regression

# **1** Introduction

nthropometric data are very important data which researchers use for different professional, academic, social and nutritive purposes. The data in use to represent anthropometric measurements are in many different formats as body dimensions vary significantly according to the population of origin. For gender classification, Lotman et al (1998) and Heinz et al (2003), could classify gender using bicromial diameters. In forensic science, Joyce and Store (1991), Wingate (1992), and Heinz at al (2003) claimed that gender of adults could be determined accurately up to 70% or more using the skeletal remains of their anthropometric measurements mainly pelvic and skull measurement. According to the argument of Heinz et al (2003), White and Churchhill (1975), Clauser at al (1977), Centre for Disease Control (CDC) office (Undated but assessed 2004), the anthropometric dataset like the one in this study could be used in ergonomic design equations if completed with data on body segment length. Aroskar and Panas (2004), Abdali at al (2004) through multiple regression analysis explored relationship in body dimensions for 507 physically active adults using anthropometric dataset which is similar the data set of this study. They also used anthropometric dataset as well as age, height and gender to predict scale weights of these individuals. There what predict scale weight best and other reasons are what motivated this study. Twelve regression models for weight were fitted to predict scale weights teenagers using their anthropometric of

measurements as measures of their body dimensions as well as their ages, heights and gender. In doing these, the dataset was categorized into three namely: all variables (all dataset of skeletal and girth measurements) as well as height. i.e. skeletal and girth body dimensions, age and gender; reduced variables (girth measurements i.e. girth body dimensions) as well as height, age and gender; reduced variables (skeletal measurements i. e. skeletal body dimensions) as well as height, age and gender.

The Pearson correlation matrix ex-arrayed the relationship in body dimensions of the teenagers. The models were fitted for boys and girls, and both. The models fitted for both sexes are the focus. Multivariate multiple linear regression method was employed to fit the models. Ordinary regression and stepwise procedures were adopted to access and select the best models.

#### 2.0 Theory of Multivariate multiple Regression Analysis

Let  $X_{i1}, X_{i2}, ..., X_{it}$  be multivariate random variables (such as skeletal (diameter) and girth circumference measurements as well as height, age and gender) assumed to be related to a response variable Y (body weight). The predictor variables of interest X determined by a mean which depends continuously on the values of the  $X_{ii}$ , i = 1, 2, ..., m, j = 1, 2, ..., t and the error

$$O_{ij}$$
 are

treated as fixed variable units. The response is considered as a random variable with behavior completely determined by the distributional assumptions, Johnson and Weichem (1992). For M responses  $Y_1, Y_2, ..., Y_m$ , with a single set of predictor variables,  $X_{i1}, X_{i2}, ..., X_{it}$ , each response follows its own regression model and we have [ see, Johnson and Weichem (1992) and Statsoft (2003)]

$$Y_{1} = \beta_{0} + \beta_{1}X_{11} + \beta_{2}X_{12} + \dots + \beta_{t}X_{1t} + \varepsilon_{1}$$

$$Y_{2} = \beta_{0} + \beta_{1}X_{21} + \beta_{2}X_{22} + \dots + \beta_{t}X_{2t} + \varepsilon_{2}$$

$$\vdots \qquad \vdots \qquad \vdots \qquad \vdots \qquad \vdots \qquad \vdots$$

$$Y_{m} = \beta_{0} + \beta_{1}X_{m1} + \beta_{2}X_{m2} + \dots + \beta_{t}X_{mt} + \varepsilon_{m}$$

The error term  $\varepsilon = [\varepsilon_1, \varepsilon_2, \dots, \varepsilon_m]'$  is a vector with the usual properties  $E(\varepsilon) = 0$ ,  $Cov(\varepsilon) = E(\varepsilon'\varepsilon) = \delta_l^2$  and  $Cov(\varepsilon_i, \varepsilon_j) = 0, i \neq j$ . The multivariate multiple linear regression model which is the form of all the models fitted in this study in matrix form is

$$\begin{bmatrix} Y_1 \\ Y_2 \\ \vdots \\ Y_m \end{bmatrix} = \begin{bmatrix} 1 & X_{11} & X_{12} & \cdots & X_{1t} \\ 1 & X_{21} & X_{22} & \cdots & X_{2t} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & X_{m1} & X_{m2} & \cdots & X_{mt} \end{bmatrix} \begin{bmatrix} \beta_1 \\ \beta_2 \\ \vdots \\ \beta_t \end{bmatrix} + \begin{bmatrix} \varepsilon_1 \\ \varepsilon_2 \\ \vdots \\ \varepsilon_m \end{bmatrix}$$
(2.1)

Since the general purpose of multiple linear regression is to quantify the relationship between overall predictor variable, so let the j<sup>th</sup> trial represent the vector of the variables (skeletal, girth measurements and other measurements of height, age and gender) as a vector of independent variables  $X_j = \begin{bmatrix} X_{j0}, & X_{j1}, & \cdots, & X_{jt} \end{bmatrix}'$ , the vector of m response variables  $Y_j = \begin{bmatrix} Y_{j1}, & Y_{j2}, & \cdots, & Y_{jm} \end{bmatrix}'$  and vector of the error term  $\varepsilon_j = \begin{bmatrix} \varepsilon_{j1} & \varepsilon_{j2}, & \cdots, & \varepsilon_{jm} \end{bmatrix}'$ . The variables of interest, body scale weight, height, age, gender and other 21 measurements from twenty one different body sites where measured and recorded sequentially are represented with the usual design matrices. It is worthy of note that the dataset of this study satisfies the assumptions of multivariate data.

2.1 Parameter Estimation

The major goal of linear regression procedure is to fit a function which predicts the response variable for given values of the predictor variables. For good fit, the choice of the regression coefficients  $\beta$  and the error variance  $6^2$  must be consistent with the data hence use of the method of least squares estimation. With the outcomes of Y, the values of the predictor variables X is of full column rank, the least squares estimate of  $\hat{\beta}_{ij}$  are exclusively determined from the observation  $\hat{Y}_i$  on the  $i^{th}$  response. So from the usual regression model

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$$Y = X\beta + \varepsilon \tag{2.2}$$

Where Y is an mxn column matrix, X is mx(t+1) matrix,  $\beta$  is a (t+1)xm vector of unknown parameters and  $\epsilon$  is the mxn vector of error term;

(2.3)

Then we have,

$$\hat{oldsymbol{eta}}_{(i)} = (X X)^{-1} X Y_{(i)} = oldsymbol{eta}$$

For any parameter  $\beta = \begin{bmatrix} b_{(1)} & \vdots & b_{(2)} & \vdots & \cdots & \vdots & b_{(t)} \end{bmatrix}$ 

With

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$$=Y - X\hat{\beta}$$
(2.4)

The error sum of squares and cross products are given as

$$\sum_{i=1}^{m} \varepsilon^{2} = \sum_{i=1}^{m} \varepsilon' \varepsilon = \sum (Y_{i} - Xb_{i})' (Y_{i} - Xb_{i})$$

The choice of  $b_{(i)}$  minimizes the  $i^{th}$ -diagonal sum of squares  $(Y_{(i)} - Xb_{(i)})'(Y_{(i)} - Xb_{(i)})$  and the trace of  $(Y - X\beta)'(Y - X\beta)$  is minimized by the choice of  $\beta$ . Using the least squares estimates  $\hat{\beta}$  matrices of the predicted models and residuals are, respectively;

$$\hat{Y} = X\hat{\beta} = X(XX)^{-1}XY$$
(2.5)

and

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$$\hat{\Sigma} = Y - \hat{Y} = \left[ I - X (X'X)^{-1} X' \right] Y$$
(2.6)

The orthogonality conditions among the residuals predicted model (values) and columns of X also hold since,  $X' [I - X (X'X)^{-1} X'] = X' - X' = 0;$  and  $X' \hat{\Sigma} = X' [I - X (X'X)^{-1} X'] Y = 0$  $\hat{Y}' \hat{\Sigma} = \hat{\beta} X [I - X (X'X)^{-1} X'] Y = 0$  since,  $Y = \hat{Y} + \hat{\varepsilon}$  *i.e.*  $Y'Y = (\hat{Y} + \hat{\Sigma})' (\hat{Y} + \hat{\Sigma}) = \hat{Y}'\hat{Y} + \hat{\Sigma}\hat{\Sigma}_{,}$ which by implication,

$$\hat{\Sigma}\hat{\Sigma}' = Y'Y - \hat{\beta}'X'X\hat{\beta}$$

We recall the best linear unbiased property of  $\hat{\beta}$ ,  $E(\hat{\beta}) = \beta$  and  $Cov(\beta_i \beta'_k) = C_{ik}^2 (X'X)^{-1}$ ; *i*, *k* = 1, 2 with residuals  $\hat{\Sigma} = [\hat{\Sigma} \odot \vdots \hat{\Sigma} \odot \vdots \hat{\Sigma} \odot \vdots \hat{\Sigma} \odot ] = Y - X\hat{\beta}$ 

$$\sum = [\sum_{(1)} : \sum_{(2)} : \cdots : \sum_{(k)}] = I - X\beta$$
  
satisfying (i)  $(\hat{\Sigma}_{ik}) = 0;$  (ii)  $E(\hat{\Sigma}\hat{\Sigma}') = \frac{n-t-1}{C_{ik}^2}$  so that

$$E\left(\frac{\hat{\Sigma}\hat{\Sigma}'}{n-t-1}\right) = C_{ik}^2 I = \Sigma$$

For multivariate multiple linear regression, the hypothesis

$$H_0: \beta_1 = \beta_2 = \dots = \beta_k = 0$$
  
$$H_1: \beta_{(i)} \neq 0, i = 1, 2, \dots, k \text{ for some } i$$

Is investigated using any of the following; F-test, t-test, Hotelling's T<sup>2</sup>-test, all of which according to Garson (2005) are special cases of Wilks Lambda test and likelihood ratio test expressed in generalized inverse form which is equivalent to Wilks Lambda Statistic (Johnson and Wichem (1992)). For details of t-test and F-test and other procedures for fitting regression models, see Neter et al (1996).

In this study, the Mallow's  $C_P$  Statistics was employed to assess the adequacy of the regression models and to select the 'best' regression equation, see Statsoft (2003) and Neter et al (1996). The Mallows  $C_P$ -statistic is of the form

$$C_{p} = \frac{RSS_{p}}{S^{2} - (n - 2p)}$$
(2.7)

Where  $RSS_P$  is the residual sum of squares from a model containing P parameter including  $\beta_0$ . The S<sup>2</sup> is the residual mean square from the largest equation postulated containing all the X<sub>(is)</sub>. It is important to recall that a regression model with how C<sub>P</sub> about equal to P is concluded to be free from lack of fit.

## 3.0 Materials and Methods

Multivariate multiple linear regression analysis was employed to explore body dimensions to predict scale weight of t = t + 1 (reflaced as the selecting the best model)

## 3.1 Data Collection and Description

In the study, locally fabricated anthropometer, plastic tape, scale weight and extended meter rule were used to collect anthropometric measurements (variables) from twenty- one body sites of six hundred and four (604) physically active teenagers within the age range of (10-20) years. The dataset consist of these anthropometric variables measurement in centimeters as well as heights (centimeters), weights (in kilogrammes), ages (in completed years), and gender of these six hundred and four teenagers. The measurement techniques and the body sites used were as suggested and prescribed by Belinke and Wilmore (1994). Nine skeletal (diameter) measurements (variables) included in the dataset were measured using the anthropometer. The skeletal variables include those measured at indicated body sites bicromial (BCRL), biiliac (BILC), bitrochanteric (BTCT), Chest (CHTD), elbow (ELBD), wrist (WRTD), knee (KNED), ankle (ANKD) diameters and chest dept (CHTP). The twelve girth (circumference) measurements (variables) used in this study are the three bony girths of the wrist (WRTGT), knee (KNEGT), ankle (ANKGT) and other nine girths measured at the sites shoulder (SHDGT), Chest (CHTGT), Waist (WATGT), Navel (NAVGT), Hip (HPGT), Thigh (THGT), Bicep (BCPGT), Forearm (FRAGT) and Calf (CAFGT), all measured with plastic tape. Each respondent had his/her age, weight, height, gender and anthropometric measurements recorded given a total of twenty-four predictor variables and one response variable (weight).

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The multivariate multiple linear regression analysis was used to analyze the data since the data are multivariate data satisfied the usual assumptions. In the normal and stepwise regression analysis procedures were used. Nine (9) different regression models were run using the normal regression procedures with statistical packages for Social sciences Version 10.0. The stepwise regression procedures were used to run three (3) different models giving a total of twelve (12) regression models. The first three models run using normal regression procedure and the last three models using stepwise selection procedure will be presented in this study. For others, the correlation matrix of the variables, plots of histograms for some of the variables, normal p-p plots of regression statistics for the models, the dataset (ANPE datast), results of parameter estimation and contribution of the variables to weight in each of the models (see, Uchegbulam, 2006).

#### 4.0 Result and Discussion

The six models discussed are presented in full

Model 1: Full model, all dataset (n=604) all variables (24) Model FMADAV.

The regression equation is:

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 \begin{split} wt &= -22.6215 + 2.8868AG - 0.9185HT + 0.8564CHTD + \\ 0.5923CHTP + 0.6540ELBD + 1.7878WRTD + 0.421KNED + \\ 0.3075ANKD + 0.3789BCRL + 0.7682BILC + 0.4015BTCT + \\ 0.1845WRTGT - 0.1727KNEGT - 0.0669ANKGT + \\ 0.2696SHDGT + 0.2971CHTGT - 0.0902WATGT + \\ 0.3064NAVGT + 0.1549HPGT + 0.1437THGT - 0.2553BCPGT + \\ 0.4591FRAGT - 0.0624CAFGT \end{split}
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With R<sup>2</sup>= 80.78%, R<sup>2</sup>- adjusted = 80.02%, S. E. = 10.015, C<sub>P</sub> = 26, F-test value =0.000, and residual mean square of 100.10. Most influential explanatory variable on the basis of their t- values are age, height, chest diameter, chest depth, elbow diameter, chest girth, forearm, wrist diameter, navel girth, biilliac and bicromial diameters.

Model 2: Full models, all dataset (604), reduced variables, (skeletal variable= 9, age and height)

Model FMADRVS:

The weight function for this model is

$$\begin{split} Wt &= 6.5914 + 2.8046AG - 0.8950HT + 1.0856CHTD + \\ 1.0316CHTP + 0.4712ELBD + 1.2082WRTD + 1.3722KNED + \\ 0.8543ANKD + 0.6233BCRL + 1.1965BILC + 0.4593BTCT \\ (4.2) \end{split}$$

WITH  $R^2=77.23\%$ ,  $R^2$ -adjusted = 76.61%, S. E. =10.79, C-P=12, F-test value= 0.000 and residual mean square of 116.45. The most contributing predictor variables on the basis of their t-test are age, height, chest diameter, chest depth, elbow diameter, knee diameter, biilliac and bicromial diameters.

Model 3: Full model, all dataset (604), reduced variables (girth variables = 12, age and height)

# Model FMADRVG

The weight function for this model is:

$$\begin{split} Wt &= 1.0895 + 3.2541AG - 0.8566HT + 1.3276WRTGT + \\ 0.5569KNEGT + 0.3479ANKGT0.3931SHDGT 0.1936CHTGT - \\ 0.0010WATGT + 0.1427NAVGT + 0.2704HPGT + 0.1410THGT + \\ 0.3239BCPGT + 0.6315FRAGT - 0.1103CAFGT \end{split}$$

With  $R^2=76.12\%$ ,  $R^2$ -adjusted = 75.56\%, S. E. 11.098,  $C_{-P}15$ , P-value=0.000 and residual mean square of 122.7166. The most influential predictor variables are age, height, forearm, hip, chest and knee girth.

The three models are all of good fit judging by their F-value and  $C_{-p}$  values. The predictor variables predicted the response variable in each case (model) with sufficient accuracy as indicated by the R<sup>2</sup> and R<sup>2</sup>-adjusted criteria. However, model FMADRVS is a better model than the model FMADRVG on the basis of R<sup>2</sup> and R<sup>2</sup>-adjusted criteria, while model FMADAV is better than both models on the same criteria. This result suggest that separating the predictor variables into skeletal variables and girth variables does not improve the quality of the regression model thereby confirming the hypothesis that scale weight is better predicted by skeletal and girth body dimensions (variables) as well as age and height.

Assessing the performance of these models obtained through ordinary regression procedure was used to run the same models and the results obtained are hereby presented.

Model 4: Full model, all dataset (n=604) all variables (24), using stepwise (spw)

Mdel SPW – FMADAV

Twelve different regression models were obtained and the optimum on the basis of  $R^2$ ,  $R^2$ -adjusted and S. E. is

$$\begin{split} Wt &= -21129 - 0.9144HT + 2.996AG + 0.483BCRL + \\ 0.216THGT + 0.498FRGT + 1.075BILC + 0.333NAVGT + \\ 0.910CHTP + 0.651CHTP + 0.5342ELBD + 0.309CHGT + \\ 2.037WRTGT \\ (4.4) \end{split}$$

Having R<sup>2</sup>=80.3%, R<sup>2</sup>-adjusted=79.9% and S.E. 10.053 with the most influential predictor variables on the basis of their t-test values are age, height, forearm girth, bicromial diameter, naval girth, chest diameter, chest depth, elbow diameter, chest girth and wrist diameter. It is important to note that the

variables dropped finally in this model were those rendered redundant by the presence or (inclusion) of some other variables which are highly correlated with them. For instance ankle diameter and wrist diameter was favoured. For the same reason, the thigh girth which is strongly correlated with bicep girth, hip girth and waist girth was favoured

Model 5: Full model, all dataset (n=604), reduced variables skeletal variables (9), age and height, using stepwise regression

#### Model SPW-FMADRVS

Eight regression functions were obtained and the optimum for the group being

Wt = 8.720 - 0.891HT + 2.952AG + 0.632BCRL + 1.457KNED +1.512BILC + 0.495ELBD + 1.004CHTP + 1.050CHTP(4.5)

With R<sup>2</sup>=76.9%, R<sup>2</sup>-adjusted=76.6%, S. E. 10.84 and residual mean square of 117.530. out of the nine (9) skeletal variables only six (6)

Table 4.1: Summary of the results

appeared in the optimum regression function. This again is as a result of strong correlation between and among body dimesions. Bitrochanteric diameter was out because of the correlation with biilliac diameter.

Model 6: Full model, all dataset, (n=604), reduced variables (girth) (12), age and height using stepwise.

#### Model SPW- FMADRVG

Twelve different regression functions were obtained and the basis of the usual criteria the optimum for this group is:

Wt = 1.134 - 0.856HT + 3.266AG0.621FRAGT + 1.503WRTGT +0.231 NAVGT + 0.415 SHDGT + 0.20 HPGT + 0.178 CHTGT +0.573KNEGT + 0.339ANKGT(4.6)

With R<sup>2</sup>=76.0%, R<sup>2</sup>-adjusted=75.5%, S. E.= 11.081 and residual mean square error = 112.780. Only eight girth variables of forearm, wrist, naval shoulder, hip, chest, knee and ankle were most influential.

Model	No of variables	R <sup>2</sup> (%)	R <sup>2</sup> -adjusted (%)	S. E.	C-p	F- value	Pr>F/sign.F
FMADAV (1)	24	80.76	80.02	10.0152	26	106.01	<0.001
FMADRVS (2)	12	77.23	76.81	10.7911	12	182.53	<0.001
FMADRVG (3)	15	76.12	75.56	11.0778	15	134.14	<0.001
Spw- FMADAV (4)	24	80.3	79.9	10.0531		200.38	0.000
SPW- FMADRVS (5)	12	76.9	76.6	10.8411	-	247.61	0.000
SPW- FMADRVG (6)	15	76.0	75.5	11.0806	-	187.27	0.000

Comparing these six models which we obtained using two different regression procedures on the whole dataset (n=604), we observed that elimination of the predictor variables considered redundant by6 the stepwise regression procedure did not improve the quality of the fit. This suggests that the predictor variables not included in the models by this procedure may not be completely uninfluential. Models 2, 3, 5 and 6 obtained with reduced predictor variables did not show any better fit either. Comparing models 1 and 4 the two full models we noticed that model 1 (FMADAV) is a better model. This suggests that the influences of all the explanatory variables are import in predicting body scale weight.

#### 4.1 Model Comparison

The six regression models obtained in this study using normal and stepwise regression procedures compared well with the mo0dels obtained using best subset and forward- backward regression selection procedures by Aroskar et al (2004) whose dataset was similar to the dataset of this study but for their age range of (21-30) years and geographical location. The R<sup>2</sup> statistic is sufficiently high for all the fitted models in this study as was the case with the models of study conducted by Aroskar et al (2004). This is an indication that the models of this study are good fits for the data, agreeing with the conclusions of Aroskar et al (2004).

Similarly, Heinz et al (2003) identified model 2 of their study as body build equation for weight for their respondents. This

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comparison is adequate because the skeletal variables are assumed to be highly consistent over years are not affected by changes in the body fat or muscle mass. Judging from model2 of

## 4.2 Conclusion

In this study, we have explored body dimensions of skeletal and girth anthropometric measurements to predict scale weight of teenagers via multiple linear and stepwise regression approach. The analysis showed that all the models are 'good fit' for the data and are adequate judging by their R<sup>2</sup>, R<sup>2</sup>- adjusted values. Also shown by the study is that reduction in the variables to skeletal or girth alone or using stepwise regression selection procedure did not improve the quality and adequacy of the fitted models. Thus body scale weight is better predicted significantly by anthropometric measurements of both skeletal and girth body dimensions as well as age, height and gender than any other group of reduced variables.

Finally, the models of this study compared well with the models of Aroskar et al (2004) in Georgia Atlanta, U. S. A. and Heinz et al (2003) in California, U. S. A.

## References

Abdali O., Vikhtor H., Paquet E. and Ritous N. Exploring Anthropometric Data through Cluster Analysis. www. Iit-iticnrc.gc.ca/iit-publications.iti/docs/NRC-46564.ppt, (2004).

Aroskar, N. and Panes A Exploring Relationships in Body Dimensions. [Regression Analysis]. Trran Projects for ISYE 644, Statistical Models and Regression Analysis. <u>www.projects</u>, amaleshpanse.corn/stat/reporte. Pdf, (2004).

Behnke A. R. and Wilmore J. H. Evaluation and Regulation of Body Build and Composition. Eaglewoods Citiffs, N. J: Prentice Hall, (1994).

Clauser, L; Tucker, P; McConville, J. Churchhill, E; Laubach, L; and Reardon, J. Anthropometry of Force Women, Report Number AMRL-TR-70-5. Aerospace Medical Research Laboratory. Wright-Patterson Air Force Base, Ohio, (1972).

**Draper, N., R. and Smith H..** Applied Regression Analysis. 3<sup>rd</sup> ed. Newyork. John Wiley and sons, **(1998)** 

Garson, G. D. GLM: MANOVA and MANCOVA. www2.chess.ncsu. edu/garson/pg765/ manova.htm, (2005).

Heinz, G., Peterson L. J. Johnson R. W and Lork C. J.. ExploringRelationshipsinBodyDimensions.JournalofStatisticsEducation:Vol.11.No.2,(2003)www.Amstat.org/publications/jse/vlln2/datasets.heinz.htm/

Johnson, R. A. and Wicherm, D. W. Applied Multivariate Statistical Analysis. 3<sup>rd</sup> ed. Prentice Hall. Inc. New Jesey. (1992) Heinz et al (2003) and model FMADRVS of this study it was observed that the girth measurements in model 2 of Heinz et al (2003) are those not affected by body fat and adult age.

**Joyce, C., and Stove** EWitness from the grave. The stories Bones Tell. Boston. MA. Little, Brown and Companys Pg.177-178, (1991).

Lohman, T. Roche, A. and Morterell, R (1988). Anthropometric Standardization Reference Manuel. Campaign II: Human Kinetics Books. Multiple Regression, Stat. soft, inc, (1994-2003).

Neter, J.; Kutner, M. N.; Nachitsterm, G, J. and Wassermen, W. Applied Linear Statistical Models. MCB/MC. Graw- Hill a division of the Mc: Graw-Hill Companies, U. S. A., (1996)

**Nwobi, F. N. and Nduka, E. C** Statistical Notes and Tables for Research. 2<sup>nd</sup> ed. Afro-Ortis Publications Nsukka, Nigeria, **(2003)**.

White, R. N. and Churchill, E. The Body Size of Soldiers: U S Army. Anthrometry - 1966 Report Number T2-51-CE (CPLSEL-94). U. S Army Natick Laboratories/ Natick, M. A., (1971).

**Wingate, A.** Science of the Crime: A writer's guide to Crimescience investigations, Cincinate, OH; Writer's Digest Books, P. 148, (1992).



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