Multivariate methods for Investigating Academic Stumbling: A Case Study Faculty of Economics and Rural Development, University of Gezira State, Sudan (2017)

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Abstract
The importance of factor analysis is to summarize many of the variables to a smaller number of variables known factors. The aim of the research to find out the academic stumble students of the Faculty of Economics and Rural Development, University of Gezira. A sample of 100 students was tested using A questionnaire form consisting of three axes applied to the area of the study. The analytical descriptive method was used and one of the methods of multiple variables (Factor Analysis) was used the Statistical Packages for Social Sciences (SPSS). The results indicate that the main weights affecting the first factor were weak motivation and desire to study and desire to work and work more than the study either affect the second factor is the disintegration of the family because of divorce or otherwise Accordingly, the third factor was affected by the weakness of the mental abilities and the difficulty in absorbing some subjects. The adoption of some faculty member’s inappropriate methods in teaching led to a gap between the needs of the student and the content of the subject. Accordingly, the research came out with the recommendations of intensifying the process of guidance in the departments and taking care of academically stumbled students.

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1. Introduction:
University education in Sudan is great importance due to cultural factors, national factors, economic and political factors this is due to the large numbers of students who join universities to improve their social and living conditions, because education is the source of income.
There is no doubt that the academic stumbling in the request of the University of the gezira is increasing in severity in the last years to the extent that the phenomenon is being talked about where the proportion of high and many researchers believe that this stumbling is due to personal reasons attributed to the student and social causes and the difficult political and economic conditions and others believe that the reason for this stumbling is due to purely administrative reasons practiced by the university right to demand in order to implement the regulations and laws. Academic stumbling has become a concrete phenomenon Many students suffer from it and must find appropriate solutions. The researcher believes that information technology may contribute to solving a large part of the underlying causes of this problem especially with regard to administrative aspects, and this study was exhausted by this study, where the researcher developed a proposed experimental scenario for dealing with this phenomenon in the case of students who have failed studying at the university of Gezira.

2. Research Question:

1- Does the level of academic failure vary among students according to the type of students?
3- Does the level of academic failure vary among students at different levels of study?

3. Research Objectives:

1- To estimate the level of academic failure vary among students according to the type of students?
2- To estimate the level of academic failure vary among students at different levels of study?

4. Importance of the Research:

It is a topic worthy of study and research, where academic stumbling has become suffers from large numbers of students, so it is considered standing on this phenomenon and its magnitude and knowledge of the factors affecting it is of great importance a concrete phenomenon It is clear in Sudanese universities. Since academic stumbling is associated with the performance of university students, this study is gaining importance,

As it focuses on university youth as a target group
Society, and it has many responsibilities in the future of the country, so it is the treatment of the performance of students.

Universities are always an urgent necessity.

This study may reveal the real reasons behind the students' academic stumbling, and therefore they are

Provide feedback to stakeholders, helping them to plan for appropriate solutions to address these causes to mitigate the phenomenon of stumbling.

5. Research Methodology:
The research mainly depends on descriptive analytical method and Factor Analysis.

5.1 Data Sources:-
Primary data is the main source of data collection. A questionnaire designed to solve the problem and include all the variables affecting (difficulties related to the academic side, Difficulties related to student personal circumstances, Difficulties related to economic and social factors).

6. Factor Analysis Model:
Multiple linear regression model:

\[ x_1 = \lambda_{11}f_1 + \cdots + \lambda_{1k}f_k + u_1 \]
\[ x_2 = \lambda_{21}f_1 + \cdots + \lambda_{2k}f_k + u_2 \]
\[ \vdots \]
\[ x_p = \lambda_{p1}f_1 + \cdots + \lambda_{pk}f_k + u_p \]

Where:
\( x = (x_1, \ldots, x_p)' \) are the observed variables (random)
\( f = (f_1, \ldots, f_k)' \) are the common factors (random)
\( u = (u_1, \ldots, u_p)' \) are called specific factors (random) \( \lambda_{ij} \) are called factor loadings (constants)

In short: \( x = \Lambda f + u \),
where \( \Lambda \) is the \( p \times k \) matrix containing the \( \lambda_{ij} \)'s.

Difference with multiple regression: common factors \( f_1, \ldots, f_k \) are unobserved.
6.1. Assumptions of Factor Analysis Model

Factor analysis is designed for interval data, although it can also be used for ordinal data (e.g. scores assigned to Likert scales). The variables used in factor analysis should be linearly related to each other. This can be checked by looking at scatter plots of pairs of variables. Obviously the variables must also be at least moderately correlated to each other, otherwise the number of factors will be almost the same as the number of original variables, which means that carrying out a factor analysis would be pointless (Cornish, 2007).

\[
\begin{align*}
E(x) &= 0 \text{ (if this is not the case, simply subtract the mean vector)} \\
E(f) &= 0, \text{ Cov}(f) = I \\
E(u) &= 0, \text{ Cov}(u_i, u_j) = 0 \text{ for } i \neq j \\
\text{Cov}(f, u) &= 0
\end{align*}
\]

(Variance of \(x_i\))

Notation:

\[
\text{Cov}(u) = \Psi = \text{diag}(\Psi_{11}, \ldots, \Psi_{kk})
\]

\[
\text{Cov}(x) = \sum \sigma_{ii} = \text{var} \sum_{j=1}^{k} \lambda_{ij}^2 + \psi_{ii}
\]

\[
\text{Var}(x_i) \text{ consists of two parts:}
\]

\[
h_i^2 = \sum_{j=1}^{k} \lambda_{ij}^2
\]

called communality of \(x_i\), represents variance of \(x_i\) that is shared with the other variables via the common factors. \(\psi_{ii}\), called the specific or unique variance, represents the variance of \(x_i\) that is not shared with the other variables.

\[
\text{Covariance matrix of } x
\]

\[
\sigma_{ii} = \text{cov}(x_i, x_j) = \sum_{j=1}^{k} \lambda_{ie} \lambda_{je}
\]

Hence, the factor models leads to

\[
\Sigma = \Lambda \Lambda^T + \Psi
\]
6.2 Analysis of Co-variances and Correlations
Since factor analysis usually works with the variances and co-variances of the observed x variables, it is sometimes referred to as “the analysis of covariance structures”. Some hint of this is apparent in equation, where the absence of an intercept term suggests that the means of the observed variables are either zero or of no direct interest. Indeed, this is typically the case in factor analysis, where the task is to learn about inter-relationships among variables rather than model the levels of each variable. Moreover, it is generally not possible to estimate both the factor loadings and intercept terms (Joreskog and Sorbom, 1979, cited in (Cornish, 2007). also Bollen, (1989),cited in (Cornish, 2007). Consequently, all the x variables and the unobserved n are presumed to have zero means, constraining any intercept term in equation to zero. In addition, for the ordinal variables frequently encountered in surveys, the latent variable approach to generating a correlation matrix posits the variances of the latent variables to be 1, making the all co-variances between the latent variables interpretable as correlations,(Cornish, 2007).

6.3 Steps in Factor Analysis:
The factor analysis model can be written algebraically as follows. If you have p variables $x_1, x_2, \ldots, x_p$ measured on a sample of n subjects, then variable i can be written as a linear combination of m factors $f_1, f_2, \ldots, f_m$ where, as explained above $m < p$. Thus,

$$x_i = \lambda_{i1}f_1 + \lambda_{i2}f_2 + \ldots + \lambda_{im}f_m + e_i \quad (11)$$

where the $\lambda_{ik}$ are the factor loadings (or scores) for variable i and $e_i$ is the part of variable $x_i$ that cannot be 'explained' by the factors.

There are three main steps in a factor analysis:
Step 1 Calculate Initial Factor Loadings
The factor loadings $\lambda$ are parameters to be estimated that tap how the unobserved factors account for the observed variables: the larger the values of $\lambda$, the more a particular variable is said to “load” on the corresponding factor. Note that the factor loadings $\lambda$ vary across survey items, but not across individuals. Put differently, items vary in the way they are explained by the underlying factors, but the relationships between underlying factors and observed responses is constant across individuals (hence the absence of an i subscript indexing $\lambda$). Note also that there
are fewer underlying factors than there are variables (p < k), consistent with the notion that like any statistical procedure, factor analysis is a device for ‘data reduction’ taking a possibly rich though unwieldy set of survey responses and summarizing them with a simpler underlying structure.(Cornish, 2007).

Calculate initial factor loadings can be done in a number of different ways:

**Principal Component Method:** As the name suggests, this method uses the method used to carry out a principal components analysis. However, the factors obtained will not actually be the principal components (although the loadings for the k\textsuperscript{th} factor will be proportional to the coefficients of the k\textsuperscript{th} principal component) (Cornish, 2007).

(Child D, 1973) in Principal components method: found the linear combination that explains the maximum variance from the X’s. This is the first factor. then find the next combination that explains the maximum proportion of the remaining variance and is orthogonal to the next factor, etc. (proceed until all variance is explained).

Let \((\lambda_i; e_i)\) be the eigenvalue-eigenvector pairs of \(\Sigma\), with \(\lambda_1 \geq \lambda_2 \ldots \lambda_p > 0\)

From the spectral theorem:

\[
\Sigma = \lambda_1 e_1 e_1'e_1 + \lambda_2 e_2 e_2'e_2 + \ldots + \lambda_p e_p e_p'e_p
\] (12)

Let \(L = \sqrt{\lambda_1 e_1} , \sqrt{\lambda_2 e_2} , \ldots , \sqrt{\lambda_p e_p}\)

Then:

\[
\Sigma = \lambda_1 e_1 e_1'e_1 + \ldots + \lambda_p e_p e_p'e_p = LL' = LL + 0
\]

Thus \(L\) is given by \(\sqrt{\lambda_i}\) times the coefficient of principles components, \(\psi = 0\)

Now, if \(\lambda_{m+1} , \lambda_{m+2} , \ldots , \lambda_p\) are small, then the first \(m\) principle components explain most \(\Sigma\).
Thus, with \( L_m = \sqrt{\lambda_1} e_1 \), \ldots, \( \sqrt{\lambda_m} \lambda e \)

\[ \sum \approx L_M L_M^\top \]

With specific factors this becomes

\[ \sum \approx L_M L_M^\top + \psi 3.43 \]

Where:

\[ \psi = \sigma_{ii} - \sum_{j=1}^{m} \lambda_{ij}^2 \]  

(13)

As estimator for the factor loadings and specific variances, we take:

\[ \tilde{\lambda}^\top = L_m = (\sqrt{\hat{\lambda}_1} \hat{e}_1 \ldots \sqrt{\hat{\lambda}_m} \hat{e}_m)3.45 \]

where \((\hat{\lambda}_i, \hat{e}_i)\) are the eigenvalue-eigenvector pairs of the sample covariance matrix \( S \), and

Where:

\[ \psi = S_{ii} - \sum_{i=1}^{m} \lambda_{ij}^2 \]

\( \psi \) : diagonal matrix

In many cases, the correlation matrix \( R \) (which is also the covariance matrix of the standardized data) is used instead of \( S \), to avoid problems related to measurements being in different scales (Thulin, 2011).

7. Data Analysis and Discussion

7.1 Introduction:

The study revealed the causes of academic stumbling for students of University of Gazira (Faculty of Economics and Rural Development).
The data on which this research was based is preliminary data and the analysis of the data was applied factor analysis of these data. This was done using the statistical packages program for social sciences (SPSS) in order to reach the results of the targeted analysis.

Table (1) Kaiser And Bartlett's Test

<table>
<thead>
<tr>
<th>KMO and Bartlett's Test</th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Kaiser-Meyer-Olkin Measure of Sampling Adequacy.</td>
<td>.666</td>
</tr>
<tr>
<td>Bartlett's Test of Sphericity</td>
<td></td>
</tr>
<tr>
<td>Approx. Chi-Square</td>
<td>884.806</td>
</tr>
<tr>
<td>Df</td>
<td>253</td>
</tr>
<tr>
<td>Sig.</td>
<td>.000</td>
</tr>
</tbody>
</table>

Source: prepared by the researcher of the outputs of the program (Spss).

Table (1) shows that the KOM is more than 0.5 which indicates an adequate sample size. The Bartlett test is test values are equal 884.80 significant level (0.000) This indicates that the test (significance) statistically significant (0.01). Thus, we conclude the original matrix relational is not the type of matrix unit therefore, the requirement to use the main components method has been met factor analysis.

7.2 Check the Correlation Matrix: -

In order to be valid for factor analysis they must produce at least some of the matrix relationships of 0.3 or higher. If we do not find in the matrix some of the relations of 0.3 or higher, we judge the matrix unfit for analysis factor analysis. Thus, we ignore the use of the main components method, the researcher find that the matrix of correlative relations contained some relations reached 0.3. It is also possible to use the main components method in analysis.
Table 2: Total Variance Explained

<table>
<thead>
<tr>
<th>Component</th>
<th>Initial Eigenvalues</th>
<th>Rotation Sums of Squared Loadings</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Total</td>
<td>% of Variance</td>
</tr>
<tr>
<td>2</td>
<td>1.961</td>
<td>10.893</td>
</tr>
<tr>
<td>3</td>
<td>1.857</td>
<td>10.315</td>
</tr>
<tr>
<td>4</td>
<td>1.471</td>
<td>8.170</td>
</tr>
<tr>
<td>5</td>
<td>1.307</td>
<td>7.263</td>
</tr>
<tr>
<td>6</td>
<td>.985</td>
<td>5.471</td>
</tr>
<tr>
<td>7</td>
<td>.803</td>
<td>4.464</td>
</tr>
<tr>
<td>8</td>
<td>.656</td>
<td>3.644</td>
</tr>
<tr>
<td>9</td>
<td>.514</td>
<td>2.854</td>
</tr>
<tr>
<td>10</td>
<td>.394</td>
<td>2.192</td>
</tr>
<tr>
<td>11</td>
<td>.377</td>
<td>2.093</td>
</tr>
<tr>
<td>12</td>
<td>.318</td>
<td>1.765</td>
</tr>
<tr>
<td>13</td>
<td>.270</td>
<td>1.499</td>
</tr>
<tr>
<td>14</td>
<td>.224</td>
<td>1.244</td>
</tr>
<tr>
<td>15</td>
<td>.190</td>
<td>1.055</td>
</tr>
<tr>
<td>16</td>
<td>.172</td>
<td>.956</td>
</tr>
</tbody>
</table>
source: prepared by the researcher of the outputs of the program (Spss).

Table (2) presents the total variance of the variable in initial solution. The results of initial solution extracted 16 factors the same number as number of variables factored. The factors extracted accounted for more than 1.0 unites of variance. The initial solution extracted 5 factors from accounting for more than 1.0 unit of variance.

**Factor 1:** The factor has an Eigen value equals (2.570) since this is greater than 1.0 it explains more variance than a single variable in fact (2.570) times as much. The percent a variance explained (2.570/16unitof variance) *100= (14.277%).

**Factor 2:** The factor has an Eigen value equals (2.299) it also greater than 1.0 and there for explains more variance than a single variable. The percent a variance explained (2.299/16 units of variance) *100= (12.773%).

**Factor 3:** The factor has an Eigen value equals (1.994) it also greater than 1.0 and there for explains more variance than a single variable. The percent a variance explained (1.994/16 units of variance) *100= (11.080%).

**Factor 4:** The factor has an Eigen value equals (1.864) it also greater than 1.0 and there for explains more variance than a single variable. The percent a variance explained (1.864 /16 units of variance) *100= (10.358 %).

**Factor 5:** The factor has an Eigen value equals (1.758) it also greater than 1.0 and there for explains more variance than a single variable. The percent a variance explained (1.758 /16 units of variance) *100= (9.766 %). The sum of the Eigen value associated with each factor (component) sum to16. 
(4.355+1.961+1.857+1.471+1.307+0.985+
0.0803+0.514+0.394+0.377+0.318+0.270+0.224+0.190+0.172)=16
The cumulative percentage of variance explained by the first factors was (58.252%).
Communalities: The quantity of communalities of variables represents the percentage of variance explained by the factors derived from these variables. The high this indicates that they explain the high proportion of the variance table no(3-3) Note that the factors explain the high proportion of variance where the lowest ratio is(.66)

University Connections and Confusion and shame in front of others.

Table (3) Communalities

<table>
<thead>
<tr>
<th></th>
<th>Initial</th>
<th>Extraction</th>
</tr>
</thead>
<tbody>
<tr>
<td>Weak direction and guidance in the process section</td>
<td>1.000</td>
<td>.835</td>
</tr>
<tr>
<td>Different school system in secondary education</td>
<td>1.000</td>
<td>.688</td>
</tr>
<tr>
<td>Some faculty members have inappropriate teaching methods</td>
<td>1.000</td>
<td>.861</td>
</tr>
<tr>
<td>gap between the needs of students and the content of the course material</td>
<td>1.000</td>
<td>.819</td>
</tr>
<tr>
<td>Lack of appropriate teaching aids in the educational process</td>
<td>1.000</td>
<td>.679</td>
</tr>
<tr>
<td>The classrooms are not suitable for lighting and ventilation</td>
<td>1.000</td>
<td>.744</td>
</tr>
<tr>
<td>Weak motivation and desire to study</td>
<td>1.000</td>
<td>.768</td>
</tr>
<tr>
<td>Desire to work and work more than study</td>
<td>1.000</td>
<td>.763</td>
</tr>
<tr>
<td>Weak self-confidence</td>
<td>1.000</td>
<td>.794</td>
</tr>
<tr>
<td>Confusion and shame in front of others</td>
<td>1.000</td>
<td>.663</td>
</tr>
<tr>
<td>Weakness in mental abilities</td>
<td>1.000</td>
<td>.781</td>
</tr>
<tr>
<td>Difficulty absorbing some subject</td>
<td>1.000</td>
<td>.773</td>
</tr>
<tr>
<td>University Connections</td>
<td>1.000</td>
<td>.660</td>
</tr>
<tr>
<td>There are differences between family members</td>
<td>1.000</td>
<td>.748</td>
</tr>
<tr>
<td>Disintegration of the family as a result of students or otherwise</td>
<td>1.000</td>
<td>.813</td>
</tr>
<tr>
<td>Weakness in the cultural level of the family</td>
<td>1.000</td>
<td>.758</td>
</tr>
</tbody>
</table>
Explanation of Factors: We find that the method of analysis of the main components aims to summarize a set of variables in a smaller number of factors, but there is no guarantee that we always get factors that can be easily explained by their associations with variables, and a solution to forms uses the method of rotation axes, under a process called axial rotation. This method aims at creating new factors of the factors that have already been obtained, in which their associations with the original variables must be distributed in such a way as to facilitate their interpretation. This means that the method of rotating the axes seeks to collect variables similar in nature in a way that facilitates interpretation. There are many methods to rotate the axes most important and most widely used varimax variation method, which aims to rotate the axes in such a way that makes the variance of the degrees of saturation of each factor as large as possible, that is to make the larger large and smaller specimens smaller than the design before the rotation. The variables are linked to the factors. By rotating the axes, the rotated component matrix is obtained, which shows the weights of the variable Loading. By examining the matrix of factors circulating in Table (3-4) we observe the following:

A- The main factors on the first factor (Student) were weak motivation and desire to study and Desire to work and work more than study.

B- The main factors on the second factor (Family), there are difference between family members, disintegration of the family because of students or otherwise, weakness in the cultural level of the family, Marriage in school years takes responsibility

C. The main factors on the third factor (Behavior of student) Weak self-confidence, Confusion and shame in front of others.

D. The main factors on the four factor (Academic System) Different school system in secondary education, Difficulty absorbing some subjects.

E. The main factors on the fifth factor (Department), Weak direction and guidance in the process section, Lack of appropriate teaching aids in the educational process.
Table (4) Factor one: -Student

<table>
<thead>
<tr>
<th>Variable</th>
<th>Variable name</th>
<th>Loading</th>
</tr>
</thead>
<tbody>
<tr>
<td>X₁</td>
<td>Weak motivation and desire to study</td>
<td>.735</td>
</tr>
<tr>
<td>X₂</td>
<td>Desire to work and work more than study</td>
<td>.812</td>
</tr>
</tbody>
</table>

\[ P_{C1} = .735 \times X₁ + .812 \times X₂ \]

source: prepared by the researcher of the outputs of the program (Spss).

Table (5) factor two: -Family

<table>
<thead>
<tr>
<th>Variable</th>
<th>Variable name</th>
<th>Loading</th>
</tr>
</thead>
<tbody>
<tr>
<td>X₁</td>
<td>there are difference between family members</td>
<td>.708</td>
</tr>
<tr>
<td>X₂</td>
<td>disintegration of the family as a result of students or otherwise</td>
<td>.774</td>
</tr>
<tr>
<td>X₃</td>
<td>weakness in the cultural el of the family</td>
<td>.638</td>
</tr>
<tr>
<td>X₄</td>
<td>Marriage in school years takes responsibility</td>
<td>.706</td>
</tr>
</tbody>
</table>

\[ P_{C2} = .708 \times X₁ + .774 \times X₂ + .638 \times X₃ + .706 \times X₄ \]

source: prepared by the researcher of the outputs of the program (Spss).

Table (6) factor three: -Behavior of student

<table>
<thead>
<tr>
<th>Variable</th>
<th>Variable name</th>
<th>Loading</th>
</tr>
</thead>
<tbody>
<tr>
<td>X₁</td>
<td>Weak self-confidence</td>
<td>.812</td>
</tr>
<tr>
<td>X₂</td>
<td>Confusion and shame in front of others</td>
<td>.743</td>
</tr>
</tbody>
</table>

\[ P_{C3} = .812 \times X₁ + .743 \times X₂ \]

source: prepared by the researcher of the outputs of the program (Spss).
Table (7) factor four: -Academic system

<table>
<thead>
<tr>
<th>Variable</th>
<th>Variable name</th>
<th>Loading</th>
</tr>
</thead>
<tbody>
<tr>
<td>X₁</td>
<td>Different school system in secondary education</td>
<td>.730</td>
</tr>
<tr>
<td>X₂</td>
<td>Difficulty absorbing some subjects</td>
<td>.801</td>
</tr>
</tbody>
</table>

source: prepared by the researcher of the outputs of the program (Spss).

\[ P_{C4} = .730X_1 + .801 X \]

Table (8) factor five: -Department

<table>
<thead>
<tr>
<th>Variable</th>
<th>Variable name</th>
<th>Loading</th>
</tr>
</thead>
<tbody>
<tr>
<td>X₁</td>
<td>Weak direction and guidance in the process section</td>
<td>.801</td>
</tr>
<tr>
<td>X₂</td>
<td>Lack of appropriate teaching aids in the educational process</td>
<td>.669</td>
</tr>
</tbody>
</table>

source: prepared by the researcher of the outputs of the program (Spss).

8. Conclusion:

The aim of the research is to use statistical methods to reach results that contribute to the knowledge of the most important factors that affect academic stumble. The three study variables were subjected to analysis using the main components method using the SPSS program. The research found follows to measure the adequacy of the sample size used in the study, use the KMO scale of .666, indicating the adequacy of the sample size.

- The Bartlett test was applied, which revealed that the correlation between the variables is significant, indicating the possibility of using the method of the main components in the analysis.
Standard Kaiser has revealed the existence of seven factors that increase its roots distinct from the correct one. The seven factors accounted for about 72.764 of the total variance. So that the first factor of about 14.277, the second factor of about 12.773, third factor about 11.080, fourth factor about 10.358, fifth factor about 9.766, sixth factor about 7.551 and seventh factor about 6.960.

To explain these factors, used style Varimax the biggest contrast method. We find that the first factor saturation variables (Weak motivation and desire to study, Desire to work and work more than study).

- The most significant variables contributing to itself in the formation of the second factor, Disintegration of the family because of students or otherwise.
- The most significant variables contributing to itself in the formation of the third factor, Weak self-confidence.
- The most significant variables contributing to itself in the formation of the four-factor difficulty absorbing some subjects.
- The most significant variables contributing to itself in the formation of the five-factor weak direction and guidance in the process section.

9. **Recommendations:**

1. Increase the number of hours of academic supervision.
2. Necessary follow-up from the family within the university to see the status of the student.
3. Considering the appropriate teaching methods by the professors.

**References:**

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4- Hassan's study(2001) the reasons of the failure of students at the Faculty of Education Zagazig University. College Journal 285. Education in Zagazig, January, No. 37, p. 197.


4.3References


2. 

3. 

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