# Factor Analytic Approach to Digital Forensic Investigation in Developing Countries 

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#### Abstract

The advancement of digital technology and the internet has immensely propelled socio-economic progress in developing countries. However, it also brought with it cybercrimes and the various levels of complexity. Further, in developing nations, precision and speed of cybercrime investigative process has become an intractable challenge in criminal jurisprudence. Therefore, the holistic traceability of digital (source, medium and target) devices used for cybercrimes requires the acceptance of digital forensics and adoption of a robust digital forensics investigative process. This research however, adopts a factor analytic approach to formulating and evaluating indices that contribute to the possible adoption of digital forensics in Nigeria. Here thirty-four indices were formulated and questionnaires were administered using purposive and simple random sampling techniques. The data obtained were analyzed by means of factor analysis by principal component using Statistical Package for Social Sciences (SPSS). The output was subjected to orthogonal rotation using varimax and four factors were resultantly extracted. The output of this research could serve as resource for further cyber security analysis.


## General Terms

Digital Forensics, Factor Analysis

## Keywords

Digital Forensics, Factor Analysis, Cyber Security, Contributory Indices.

## 1. INTRODUCTION

Securing the cyber space has become very indispensable with the spread of information technology and digital application. The wave of cyber-attacks, which now cuts across all spheres of human-computer interaction has resulted in significant financial and public relations losses for businesses and governments. Personal computers and other digital assets are being used more frequently in both homes and businesses, which has led to a spike in high-tech crimes. To curb the above, Nigerian government came up with an act on the $15^{\text {th }}$ of May, 2015: The Cybercrime (Prevention \& Prohibition) Act, 2015. This was enacted for the purpose of prohibiting, preventing, detecting, responding, investigating and prosecution of cybercrimes and for other related matters, 2015. The act sees any form of vandalisation or crime against critical national information infrastructure, unlawful access to computer systems, cyber grooming, and the likes as crimes which are punishable.

The increase in cyberattack and their related effects all together provide the basis for adopting systematic incident response procedures. Responding to high-tech crimes and carrying out efficient incident response operations require a methodical approach built on a reliable forensic exercise. As this approach is well adopted in developing nations, there is the need to fully incorporate it into the legal jurisprudence of developing nations
such as Nigeria.
Cybersecurity is concerned with the defense of digital assets and the preservation of data. Likewise the design, development, implementation, and management of several policies, frameworks, and strategies that direct the protection of data against unauthorized access and illegal modifications are included.

Digital forensics is a relatively new field. Derived as a synonym for computer forensics, its definition has expanded to include the forensics of all digital technology. Whereas computer forensics is defined as "the collection of technique and tools used to find evidence in a computer" [4]. Digital forensics has been defined as "the use of scientifically derived and proven methods towards the preservation, collection, validation, identification, analysis, interpretation, documentation, and presentation of digital evidence derived from digital sources for the purpose of facilitation or furthering the reconstruction of events found to be criminal, or helping to anticipate unauthorized actions shown to be disruptive to planned operations" [5]. Digital forensics, "is the discipline that combines elements of law and computer science to collect and analyze data from computer systems, networks, wireless communications, and storage devices in a way that is admissible as evidence in a court of law". The word forensics denotes use in law/courts thus signifying digital forensics as a process carried out ultimately to acquire evidence that may be used in a court of law. The goal of digital forensics, as simply stated, "is to identify digital evidence for an investigation". The fact that digital forensics has a legal connotation cannot be overemphasized [10].
[10] stated that digital forensics has become prevalent because law enforcement recognizes that modern day life includes a variety of digital devices that can be exploited for criminal activity, not just computer systems. While computer forensics tends to focus on specific methods for extracting evidence from a particular platform, digital forensics must be modeled such that it can encompass all types of digital devices, including future digital technologies. Unfortunately, here in Nigeria, there does not exist a standard or consistent digital forensics methodology, but rather a set of procedures and tools built from the experience of law enforcement, system administrators, and hackers.

## 2. RELATED WORKS

The authors in [9] explore the development of digital forensics process, compare and contrasts four particular forensic methodologies and finally proposes an abstract model of the digital forensic procedure. The research was simulated with some steps as "pre-incident preparation, detection of incidents, initial response strategy formulation, duplication, investigation, security measure implementation, network monitoring, recovery, reporting and follow-up using platform such as Windows NT/2000, UNIX and Cisco Routers. The scheme
created consistent and standardized framework for digital forensic development and also identifies the need for specific technology-dependent tools while providing insight from previously defined tools of same category. In [10] solution was produced to issues surrounding digital evidence acquisition and subsequent presentation in court and outlines guidelines for making this type of evidence more robust when presented in court. This research addressed intricate issues of the digital forensics process and lays the foundations of a frameworks that will accurately and rigorously address the multidimensional nature of the field. A process model for digital investigation is defined using the theories and techniques from the physical investigation world. While digital investigation. This model allows technical requirement for each phase to be developed and for the interaction between physical and digital investigation to be identified. It is abstract enough that it can be applied to both law enforcement and corporate scenarios [3].

## 3. RESEARCH METHODOLOGY

The contributing indices for digital forensics was derived and questionnaires were distributed to selected respondents who are knowledgeable about subject topic and are aware of the need for standardization in the area. There was a guarantee of the privacy of the personal data supplied by respondents. Age, gender, the highest academic degree had, and occupation are just a few of the biographical details that respondents submitted. To help the respondents comprehend the questions and give accurate replies, some questions in the questionnaire that contained technical words were explained.
The computer model of factor analysis by the principal components of the contributory indices to Digital Forensic Investigation Process was formulated and statistical package for social sciences (SPSS) was used in the analysis of the factors that contribute to digital forensic.
Factor analysis is a statistical method used to describe variability among observed, correlated variables in terms of a potentially lower number of unobserved variables called factors. It is used mostly for data reduction purposes to get a small set of variables from a large set of variables and also to create indices with variables that measure similar things. It is also used for summarization and also for testing the validity of a test or scale. The inter-correlation between variables was determined while conducting factor analysis by using the correlate procedure to create a correlation matrix of all variables.


Figure. 1: Simple path for factor analytic model (The Common Factor Model)

Figure 1 shows a simple path diagram for factor analysis model. Each observed response (Measure 1 through Measure 5) is influenced partially by the fundamental common factors (Factor 1 and Factor 2) and partially by the underlying factors (E1 through E5). Factor analysis by principal components of the data obtained through survey has been implemented in [2]. The mathematical model of the evaluation of contributory indices of digital forensics using the factor analysis by principal components is expressed as shown in the equation 1:
$Y_{j}=\sum_{k=1}^{m} a_{j, k} X_{k} \quad k=1,2,3, \ldots \ldots, m$

Where:
$Y_{j}$ is the $j^{\text {th }}$ respondent, $a_{j, k}$ is the assessment of the $\mathrm{k}^{\text {th }}$ variable by $\mathrm{j}^{\text {th }}$ respondent, and $\mathrm{X}_{\mathrm{k}}$ is the $\mathrm{k}^{\text {th }}$ decision.

The Descriptive statistics define the mean, standard deviation and number of respondents $(\mathrm{N})$ who participated in the survey. The correlation gives the correlation coefficients between a single variable and every other variables in the analysis. The correlation matrix contains 1 . The correlation coefficients above and below the principal diagonal are the same.
Total Variance Explained shows all the factors extractable from the analysis along with their eigenvalues, the percent of variance attributable to each factor and the cumulative variance of the factor. The first principal component (scaled eigenvector) by definition is the one that explains the largest part of the total variance.

The scree plot is a graph of the eigenvalues against all the factors. It is useful for determining how many factors to retain. For each principal component, the corresponding eigenvalue is plotted on the $y$-axis. The display of an elbow at a given value on the x -axis indicates a higher order principal component that shows a decreasing amount of additional variance.
The Component Matrix shows the loadings of all the variables on the factors extracted. The Rotated Component Matrix is aimed at reducing the number of factors on which the variables under investigation have higher loadings. It does not change anything, buy makes the interpretation of the analysis easier.

### 3.2 Data Survey and Collection

To examine relationships and outline data stored from the questionnaire, a well-structured questionnaire is created. With the aid of a well-structured questionnaire to evaluate correlations acquired from the questionnaire, the data that would be used in creating the computer model based on the developed contributing indices would be obtained.

Upon accumulating the primary data required for the research project, the outcomes were compiled. As the research employed questionnaires with adequate and reliable information, this study has a descriptive research design. It was designed with the intention of gathering accurate and sufficient data while sampling the views of different respondents in academic environment. In order to get responses from the respondents, a structured questionnaire was used. The respondents were given the survey, which was then collected.

This survey was carried out in a number of locations. In view of the large number of local governments, states, institutions of higher learning, law firms \& security outfits, certain classes of respondents were identified during this survey and used in this
sampling instruments for the purpose of the research. The classes of respondents are: Staff of institutions of higher learning, students of institutions of higher learning, staffs and clients of IT firms, legal practitioners and administrative staff at law court, and security personnel.

## 4. RESULTS AND INTERPRETATION

The respondents filled the questionnaire in accordance to their understanding with each particular index. A six-point scale of 'very high', 'high', 'average', 'low', 'very low' and 'undecided'. Two Hundred questionnaires were distributed, and same two hundred (200) were returned completely filled. The analysis of gender and age incidence of perpetrators are presented in table $1,2,3,4,5,6$ and 7 respectively.

| Table 1: Male Perpetrators |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | Freque ncy | Perc ent | Valid <br> Perc <br> ent | Cumulat ive Percent |
| Val id | Very High | 142 | 71.0 | 71.0 | 71.0 |
|  | High | 23 | 11.5 | 11.5 | 82.5 |
|  | Averag <br> e | 16 | 8.0 | 8.0 | 90.5 |
|  | Low | 5 | 2.5 | 2.5 | 93.0 |
|  | Very Low | 1 | . 5 | . 5 | 93.5 |
|  | Undeci ded | 13 | 6.5 | 6.5 | 100.0 |
|  | Total | 200 | $\begin{array}{r} 100 . \\ 0 \\ \hline \end{array}$ | 100. |  |


| Table 2: Female Perpetrators |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | Freque ncy | Perc ent | Valid <br> Perc <br> ent | Cumulat ive Percent |
| Val id | Very High | 12 | 6.0 | 6.0 | 6.0 |
|  | High | 32 | 16.0 | 16.0 | 22.0 |
|  | Averag <br> e | 65 | 32.5 | 32.5 | 54.5 |
|  | Low | 51 | 25.5 | 25.5 | 80.0 |
|  | Very Low | 15 | 7.5 | 7.5 | 87.5 |
|  | Undeci ded | 25 | 12.5 | 12.5 | 100.0 |
|  | Total | 200 | $\begin{array}{r} 100 . \\ 0 \end{array}$ | $\begin{array}{r} 100 . \\ 0 \\ \hline \end{array}$ |  |


| Table 3: Range_2_to_11 |  |  |  |  |  |  |
| :---: | :--- | ---: | ---: | ---: | ---: | :---: |
| Val <br> id | Freque <br> ncy | Perc <br> ent | Valid <br> Perc <br> ent | Cumulat <br> ive <br> Percent |  |  |
|  | Averag <br> e | 6 | 3.0 | 3.0 | 3.0 |  |
|  | Low | 26 | 13.0 | 13.0 | 17.0 |  |
|  | Very <br> Low | 78 | 39.0 | 39.0 | 56.0 |  |
|  | Undeci <br> ded | 88 | 44.0 | 44.0 | 100.0 |  |
|  | Total | 200 | 100. | 100. |  |  |
|  | 0 | 0 |  |  |  |  |


| Table 4: Range_12_to_17 |  |  |  |  |  |  |
| :--- | :--- | :--- | :---: | :---: | :---: | :---: |
|  |  | Freque <br> ncy | Perc <br> ent | Valid <br> Perc <br> ent | Cumulat <br> ive <br> Percent |  |
| Val <br> id | Very <br> High | 9 | 4.5 | 4.5 | 4.5 |  |


|  | High | 18 | 9.0 | 9.0 | 13.5 |
| :---: | ---: | ---: | ---: | ---: | ---: |
|  | Averag <br> e | 52 | 26.0 | 26.0 | 39.5 |
|  | Low | 64 | 32.0 | 32.0 | 71.5 |
|  | Very <br> Low | 35 | 17.5 | 17.5 | 89.0 |
|  | Undeci <br> ded | 22 | 11.0 | 11.0 | 100.0 |
|  | Total | 200 | 100. <br> 0 | 100. <br> 0 |  |


| Table 5: Range_18_to_25 |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | Frequenc y | Perce nt | Valid Perce nt | Cumu <br> lative <br> Perce <br> nt |
| $\begin{aligned} & \hline \text { V } \\ & \text { ali } \\ & \text { d } \end{aligned}$ | Very High | 115 | 57.5 | 57.5 | 57.5 |
|  | High | 39 | 19.5 | 19.5 | 77.0 |
|  | Average | 28 | 14.0 | 14.0 | 91.0 |
|  | Low | 2 | 1.0 | 1.0 | 92.0 |
|  | Very Low | 1 | . 5 | . 5 | 92.5 |
|  | Undecide d | 15 | 7.5 | 7.5 | 100.0 |
|  | Total | 200 | 100.0 | 100.0 |  |


| Table 6: Range_26_to_45 |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | Frequency | Percent | Valid <br> Perce <br> nt | Cumula tive Percent |
| Valid | Very High | 97 | 48.5 | 48.5 | 48.5 |
|  | High | 53 | 26.5 | 26.5 | 75.0 |
|  | Aver age | 26 | 13.0 | 13.0 | 88.0 |
|  | Low | 3 | 1.5 | 1.5 | 89.5 |
|  | Very Low | 1 | . 5 | . 5 | 90.0 |
|  | Unde cided | 20 | 10.0 | 10.0 | 100.0 |
|  | Total | 200 | 100.0 | 100.0 |  |


| Table 7: Range_Above_45 |  |  |  |  |  |  |
| :--- | :--- | ---: | ---: | ---: | ---: | :---: |
|  | Frequency | Percent | Valid <br> Percent | Cumula <br> tive <br> Percent |  |  |
|  | Very High | 14 | 7.0 | 7.0 | 7.0 |  |
|  | High | 32 | 16.0 | 16.0 | 23.0 |  |
|  | Average | 49 | 24.5 | 24.5 | 47.5 |  |
|  | Low | 36 | 18.0 | 18.0 | 65.5 |  |
|  | Very Low | 27 | 13.5 | 13.5 | 79.0 |  |
|  | Undecided | 42 | 21.0 | 21.0 | 100.0 |  |
|  | Total | 200 | 100.0 | 100.0 |  |  |

Table 8: Descriptive Statistics

|  | N | Mean | Std. Deviation |
| :--- | ---: | ---: | ---: |
| DEIP | 200 | 2.39 | 1.366 |
| DECV | 200 | 2.62 | 1.340 |
| SDDSM | 200 | 2.29 | 1.405 |
| RAPSM | 200 | 2.12 | 1.368 |
| ADCDD | 200 | 2.52 | 1.396 |
| INTH | 200 | 2.33 | 1.284 |
| EIAM | 200 | 2.31 | 1.277 |
| INBIN | 200 | 2.45 | 1.318 |
| ONFR | 200 | 2.20 | 1.410 |
| INES | 200 | 2.35 | 1.403 |
| HACK | 200 | 2.07 | 1.364 |
| SOENG | 200 | 2.24 | 1.520 |
| TERCT | 200 | 2.53 | 1.318 |


| CONLI | 200 | 2.82 | 1.219 |
| :--- | :--- | :--- | :--- |
| COMLI | 200 | 2.66 | 1.412 |
| NAPDF | 200 | 2.51 | 1.330 |
| LFDF | 200 | 2.78 | 1.216 |
| RFDF | 200 | 2.77 | 1.242 |
| IFDF | 200 | 2.74 | 1.273 |
| IMDF | 200 | 2.77 | 1.298 |
| PADF | 200 | 2.97 | 1.277 |
| POLWI | 200 | 2.85 | 1.243 |
| ICDF | 200 | 3.03 | 1.171 |
| FRDR | 200 | 2.92 | 1.305 |
| DFPAPM | 200 | 2.85 | 1.283 |
| DFRACM | 200 | 2.93 | 1.278 |
| PPAC | 200 | 3.10 | 1.210 |
| ADFT | 200 | 3.00 | 1.339 |
| RTDFP | 200 | 3.12 | 1.312 |
| ECSM | 200 | 3.03 | 1.272 |
| TLRR | 200 | 3.18 | 2.466 |
| TLIN | 200 | 2.94 | 1.302 |
| TLPE | 200 | 2.84 | 1.271 |
| DBDFAI | 200 | 2.75 | 1.318 |

The descriptive statistics of the data collected is presented in Table 8. The table shows the mean and standard deviation of the assessment of each of the contributory indices to digital forensics by the respondents. It is inferred from the mean that Hacking is the prevalent index that could likely contribute to or require digital Forensics. It has the highest mean of 2.07

The SPSS generates the correlation matrix as a single file shown in Appendix 2. The correlation between a variable and itself is always 1 , hence the principal diagonal of the correlation matrix contains 1 . The correlation coefficients above and below the principal diagonal are the same. The determinant of the correlation matrix is given as $7.209 \mathrm{E}-16$. The KMO test performed in this analysis produces a measure of 0.940 . Bartlett's test produces an $\mathrm{X}^{2}$ of 6514.144 with a significant level of 0.000 . The significant level confirms the adequacy of the sample population. The result obtained from the two tests (Bartlett's test of Sphericity and KMO test) indicate the suitability of the application of factor analysis as well. Table 9 indicates the communalities of variables, which ranges from 0 to 1 . The table shows that the communalities of 'Delayed in Investigation Process 'and 'Delayed in Court Verdict' are 0.739 and 0.723 respectively. This implies that $73.90 \%$ of the variance in 'Delayed in Investigation Process' can be explained by the extracted factors while the remaining $26.10 \%$ is attributed to extraneous factors. Similarly, $72.30 \%$ of the variance in 'Delayed in Court Verdict' can be explained by the extracted factors, while the remaining $27.70 \%$ is attributed to extraneous factors. The factor 'Time Lag in Investigation' has the highest value of communality with over $80 \%$ of the variance while 'Time Lag for Rapid Response' has the smallest value of communality with $47.5 \%$ of the variance. In addition, the factors with small values such as 'Time Lag for Rapid Response' (0.475), Social Engineering (0.575), Documented breakthroughs in Digital Forensic aided Investigation (0.579), Online Fraud (0.632); should be dropped from the analysis. Table 10 consists of the Component (Factor) Matrix table which interprets the components. It shows the loadings of the contributory factors on the four (4) components extracted using Principal Component Analysis. The higher the absolute value of the loadings, the more the factor contributes to the variable. The gap on the table represents loadings that are less than 0.3 were suppressed. In Factor 1 all the variables were loaded, factor 2 , only 13 variables were not loaded, factor 3 , only 5 variables were loaded and in factor 4 , only 3 variables were loaded.

Table 9: Communalities of Variables

|  | Initial | Extraction |  | Initial | Extraction |
| :---: | :---: | :---: | :---: | :---: | :---: |
| DEIP | 1.000 | . 739 | ADFT | 1.000 | . 698 |
| DECV | 1.000 | . 723 | RTDFP | 1.000 | . 754 |
| SDDSM | 1.000 | . 663 | ECSM | 1.000 | . 692 |
| RAPSM | 1.000 | . 689 | TLRR | 1.000 | . 475 |
| ADCDD | 1.000 | . 703 | TLIN | 1.000 | . 801 |
| INTH | 1.000 | . 673 |  |  |  |
| EIAM | 1.000 | . 772 | TLPE | 1.000 | . 719 |
| INBIN | 1.000 | . 723 | DBDFAI | 1.000 | . 579 |
| ONFR | 1.000 | . 632 | PPAC | 1.000 | . 751 |
| INES | 1.000 | . 645 |  |  |  |
| HACK | 1.000 | . 719 |  |  |  |
| SOENG | 1.000 | . 575 |  |  |  |
| TERCT | 1.000 | . 658 |  |  |  |
| CONLI | 1.000 | . 658 |  |  |  |
| COMLI | 1.000 | . 619 |  |  |  |
| NAPDF | 1.000 | . 679 |  |  |  |
| LFDF | 1.000 | . 742 |  |  |  |
| RFDF | 1.000 | . 712 |  |  |  |
| IFDF | 1.000 | . 718 |  |  |  |
| IMDF | 1.000 | . 772 |  |  |  |
| PADF | 1.000 | . 783 |  |  |  |
| POLWI | 1.000 | . 659 |  |  |  |
| ICDF | 1.000 | . 693 |  |  |  |
| FRDR | 1.000 | . 718 |  |  |  |
| DFPAPM | 1.000 | . 782 |  |  |  |
| DFRACM | 1.000 | . 737 |  |  |  |

Table 10: Component Factor Matrix

|  | 1 | 2 | 3 | 4 |
| :---: | :---: | :---: | :---: | :---: |
| DEIP | . 682 |  |  | -. 415 |
| DECV | . 691 |  |  | -. 390 |
| SDDSM | . 684 | . 366 |  |  |
| RAPSM | . 683 | . 446 |  |  |
| ADCDD | . 705 | . 414 |  |  |
| INTH | . 719 | . 388 |  |  |
| EIAM | . 747 | . 428 |  |  |
| INBIN | . 680 | . 490 |  |  |
| ONFR | . 684 | . 396 |  |  |
| INES | . 658 | . 452 |  |  |
| HACK | . 785 | . 321 |  |  |
| SOENG | . 688 |  |  |  |
| TERCT | . 701 | . 395 |  |  |
| CONLI | . 662 | . 400 |  |  |
| COMLI | . 578 |  |  | .437 |
| NAPDF | . 775 |  |  |  |
| LFDF | . 793 |  |  |  |
| RFDF | . 806 |  |  |  |
| IFDF | . 803 |  |  |  |
| IMDF | . 747 | -. 331 | -. 301 |  |
| PADF | . 729 | -. 321 | -. 323 |  |
| POLWI | . 739 |  |  |  |
| ICDF | . 763 | -. 316 |  |  |
| FRDR | . 754 | -. 380 |  |  |
| DFPAPM | . 782 | -. 389 |  |  |
| DFRACM | . 759 | -. 371 |  |  |
| PPAC | . 760 | -. 413 |  |  |
| ADFT | . 740 | -. 378 |  |  |
| RTDFP | . 695 | -. 448 |  |  |
| ECSM | . 721 | -. 360 |  |  |
| TLRR | . 378 |  | . 540 |  |
| TLIN | . 693 |  | . 523 |  |
| TLPE | . 746 |  | . 338 |  |
| DBDFAI | . 714 |  |  |  |

The Rotated Component (Factor) Matrix is the next analysis that was performed. The method of Extraction is the Principal Component Analysis. In order to obtain meaningful representation of variables, the resulted principal component is rotated by orthogonal transformation by varimax, quartimax, equamax, and promax. However, the method chosen for the analysis is Rotated Component (Factor) Matrix using Varimax.

Table 11: Rotated Component Matrix using Varimax

|  | 1 | 2 | 3 | 4 |
| :---: | :---: | :---: | :---: | :---: |
| DEIP |  | . 599 |  | . 536 |
| DECV |  | . 614 |  | . 507 |
| SDDSM |  | . 682 |  | . 355 |
| RAPSM |  | . 771 |  |  |
| ADCDD |  | . 763 |  |  |
| INTH |  | . 761 |  |  |
| EIAM |  | . 828 |  |  |
| INBIN |  | . 824 |  |  |
| ONFR |  | . 747 |  |  |
| INES |  | . 767 |  |  |
| HACK | . 345 | . 747 |  |  |
| SOENG | . 311 | . 595 |  | . 320 |
| TERCT |  | . 764 |  |  |
| CONLI |  | . 757 |  |  |
| COMLI |  | . 624 | . 304 | -. 314 |
| NAPDF | . 701 | . 388 |  |  |
| LFDF | . 667 | . 432 |  |  |
| RFDF | . 681 | . 481 |  |  |
| IFDF | . 755 | . 348 |  |  |
| IMDF | . 828 |  |  |  |
| PADF | . 813 |  |  |  |
| POLWI | . 735 | . 337 |  |  |
| ICDF | . 729 |  | . 302 |  |
| FRDR | . 783 |  |  |  |
| DFPAPM | . 837 |  |  |  |
| DFRACM | . 815 |  |  |  |
| PPAC | . 798 |  |  |  |
| ADFT | . 753 |  |  |  |
| RTDFP | . 688 |  | . 470 |  |
| ECSM | . 678 |  | . 440 |  |
| TLRR |  |  | . 655 |  |
| TLIN | . 444 |  | . 726 |  |
| TLPE | . 523 | . 302 | . 537 |  |
| DBDFAI | . 503 | . 366 |  | . 333 |

The interpretation of table 11 is as follow
Factor 1- National Policy/Legislature Framework on Digital Forensics Investigation and Public-Private-People Partnership, loads on

1. National Policy on Digital Forensics
2. Legislature's Framework on Digital Forensics
3. Regulatory Framework on Digital Forensics
4. Institutional Framework on Digital Forensics
5. Implementation of Digital Forensics
6. Political will
7. Industry Contribution to Digital Forensics
8. Funding of Research on Digital Forensics
9. Digital Forensic Pro-Active (Preventive) Measures
10. Digital Forensic Re-Active (Curative) Measures
11. Public/Private Agency Collaboration
12. Public awareness of digital forensics
13. Availability of digital forensics tools
14. Documented breakthroughs in digital forensic aided investigation

Factor 2- Motive for Digital Forensics Investigation loads on

1. Delayed Investigation Process
2. Delayed Court Verdict
3. Rapid Growth of social media
4. Accidental or Deliberate Company Data Disclosure
5. Intellectual Theft
6. Employee Internet Abuse or Misuse
7. Incident or Breach Investigation
8. Hacking
9. Online Fraud
10. Industrial Espionage
11. Terrorism (Cyber Terrorism)
12. Conventional Literacy
13. Computer Literacy

Factor 3- Security Agencies loads on

1. Regular training of digital forensics personnel
2. Effective crime scene management
3. Time lag for rapid response
4. Time lag for presentation of evidence
5. Time lag for investigation

Factor 4- Motive for Digital Forensics Investigation loads on

1. Stolen Digital Devices such as Smartphones
2. Social Engineering

Appendix 1 'Total Variance Explained’ which shows how much of the total variance of the observed variables is explained by each of the principal components is presented in appendix 1. The extraction method is Principal Component Analysis. The first principal component (scaled eigenvalue) by definition explains the largest part of the total variance. It has a variance (eigenvalue) of 17.5 ; this accounts for $51.417 \%$ of the total variance. The second principal component has a variance of 3.8 and accounts for a further $11.308 \%$ of the variance and so on. The "Cumulative \%" column of the table tells us how much of the total variance can be accounted for by the first $K$ components together. $69.58 \%$ of the extracted (4) factors contribute to the Digital Forensics based on the views of the respondents. The remaining $30.42 \%$ is considered to be the contribution of extraneous factors.

Figure 2 (below) shows the Scree Plot used in analysis. The display of an elbow at a given value on the $x$-axis indicates a higher order principal component that shows a decreasing amount of additional variance. There is a marked decrease in downward slope after the sixth principal component, although its eigenvalue is greater than 1 .


Fig. 2: Scree Plot

## 6. CONCLUSION

On the basis of the description analysis based on the gender incidence of perpetrators from the respondents, most perpetrators are much likely to be of the male gender based on the mean (1.70), while perpetrators with age range 18-25 and $26-45$, are much likely to be involved. The elements influencing digital forensics in Nigeria were identified using a factor analytic technique, which was reported in this research. The covariances and correlations between the factors are also noted. Four contributory variables were also extracted. The outcome of this research is solely due to the contributing variables that were developed and utilized in the assessment of the indicators for the use of Digital Forensics Investigation Process in Nigeria. This research focuses on the representation of digital forensics in Nigeria. It could serve as resource for further research in digital forensics (DF) process.

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## APPENDIX A: TOTAL VARIANCE EXPLAINED

| Component | Initial Eigenvalues |  |  | Extraction Sums of Squared Loadings |  |  | Rotation Sums of Squared Loadings |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Total | $\begin{gathered} \hline \% \text { of } \\ \text { Variance } \end{gathered}$ | Cumulative \% | Total | $\%$ of Variance | Cumulative \% | Total | $\%$ of Variance | Cumulative $\%$ |
| 1 | 17.482 | 51.417 | 51.417 | 17.482 | 51.417 | 51.417 | 10.18 | 29.949 | 29.949 |
| 2 | 3.845 | 11.308 | 62.725 | 3.845 | 11.308 | 62.725 | 3 | 27.831 | 57.780 |
| 3 | 1.243 | 3.656 | 66.382 | 1.243 | 3.656 | 66.382 | 9.463 | 7.562 | 65.342 |
| 4 | 1.089 | 3.204 | 69.586 | 1.089 | 3.204 | 69.586 | 2.571 | 4.244 | 69.586 |
| 5 | . 829 | 2.439 | 72.025 |  |  |  | 1.443 |  |  |
| 6 | . 801 | 2.356 | 74.382 |  |  |  |  |  |  |
| 7 | . 756 | 2.225 | 76.606 |  |  |  |  |  |  |
| 8 | . 734 | 2.158 | 78.765 |  |  |  |  |  |  |
| 9 | . 689 | 2.028 | 80.792 |  |  |  |  |  |  |
| 10 | . 619 | 1.822 | 82.614 |  |  |  |  |  |  |
| 11 | . 531 | 1.563 | 84.176 |  |  |  |  |  |  |
| 12 | . 483 | 1.421 | 85.597 |  |  |  |  |  |  |
| 13 | . 463 | 1.362 | 86.959 |  |  |  |  |  |  |
| 14 | . 424 | 1.247 | 88.206 |  |  |  |  |  |  |
| 15 | . 390 | 1.146 | 89.353 |  |  |  |  |  |  |
| 16 | . 372 | 1.094 | 90.447 |  |  |  |  |  |  |
| 17 | . 360 | 1.059 | 91.506 |  |  |  |  |  |  |
| 18 | . 309 | . 910 | 92.416 |  |  |  |  |  |  |
| 19 | . 285 | . 838 | 93.254 |  |  |  |  |  |  |
| 20 | . 261 | . 768 | 94.021 |  |  |  |  |  |  |
| 21 | . 249 | . 732 | 94.753 |  |  |  |  |  |  |
| 22 | . 222 | . 654 | 95.407 |  |  |  |  |  |  |
| 23 | . 193 | . 568 | 95.974 |  |  |  |  |  |  |
| 24 | . 182 | . 535 | 96.509 |  |  |  |  |  |  |
| 25 | . 173 | . 509 | 97.018 |  |  |  |  |  |  |
| 26 | . 160 | . 471 | 97.489 |  |  |  |  |  |  |
| 27 | . 131 | . 385 | 97.874 |  |  |  |  |  |  |
| 28 | . 123 | . 362 | 98.236 |  |  |  |  |  |  |
| 29 | . 114 | . 336 | 98.572 |  |  |  |  |  |  |
| 30 | . 111 | . 326 | 98.898 |  |  |  |  |  |  |
| 31 | . 103 | . 303 | 99.201 |  |  |  |  |  |  |
| 32 | . 101 | . 296 | 99.497 |  |  |  |  |  |  |
| 33 | . 087 | . 257 | 99.754 |  |  |  |  |  |  |
| 34 | . 083 | . 246 | 100.000 |  |  |  |  |  |  |

## APPENDIX B: CORRELATION MATRIX OF VARIABLES

Correlation Matrix

|  | DEIP | DECV | SDDSM | RAPSM | ADCDD | INTH | EIAM | INBIN | ONFR | INES |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| DEIP | 1 | .786 | . 618 | . 581 | . 603 | . $557{ }^{* *}$ | . $595{ }^{* *}$ | . $506{ }^{* *}$ | . $478{ }^{* *}$ | . $522^{* *}$ |
| DECV | . $786{ }^{* *}$ | 1 | . $632^{* *}$ | . $592^{* *}$ | .630** | . $538{ }^{* *}$ | . $576{ }^{* *}$ | . $539^{* *}$ | . $572{ }^{* *}$ | . 550 ** |
| SDDSM | . $618^{* *}$ | . 632 | 1 | . $574{ }^{* *}$ | . $611^{* *}$ | . $599{ }^{* *}$ | . $545{ }^{* *}$ | . $579 * *$ | . $566{ }^{* *}$ | . $607^{* *}$ |
| RAPSM | . $581{ }^{* *}$ | . $5922^{* *}$ | . $574{ }^{* *}$ | 1 | .645** | . $648{ }^{* *}$ | . $667{ }^{* *}$ | .685** | . $710{ }^{* *}$ | . $623{ }^{* *}$ |
| ADCDD | . $603{ }^{* *}$ | . 630 ** | .611** | . $645^{* *}$ | 1 | . $708^{* *}$ | . $672^{* *}$ | .722** | . $673{ }^{* *}$ | . $587{ }^{* *}$ |
| INTH | . $557{ }^{* *}$ | . $538^{* *}$ | . $599{ }^{* *}$ | . $648{ }^{* *}$ | . $708^{* *}$ | 1 | . $695{ }^{* *}$ | . $681{ }^{* *}$ | . $669^{* *}$ | . $641^{* *}$ |
| EIAM | .595** | . $576{ }^{* *}$ | . $545^{* *}$ | . $667{ }^{* *}$ | .672** | . $695{ }^{* *}$ | 1 | . $719^{* *}$ | .648** | .695** |
| INBIN | .506** | . $539 * *$ | . $579{ }^{* *}$ | . $685{ }^{* *}$ | . $722^{* *}$ | . $681{ }^{* *}$ | . $719^{* *}$ | 1 | . $626^{* *}$ | .654** |
| ONFR | . $478{ }^{* *}$ | . $572 * *$ | . $566{ }^{* *}$ | . $710^{* *}$ | . $673{ }^{* *}$ | .669** | . $648^{* *}$ | . $626^{* *}$ | 1 | . $522^{* *}$ |
| INES | . $522^{* *}$ | . 550 ** | . $607^{* *}$ | .623** | . $587{ }^{* *}$ | . $641^{* *}$ | .695** | . $654 * *$ | . $522^{* *}$ | 1 |
| HACK | . $626{ }^{* *}$ | . 591 ** | .643** | . $672^{* *}$ | .671** | . $668{ }^{* *}$ | .739** | . $677^{* *}$ | . $704^{* *}$ | . $587{ }^{* *}$ |
| SOENG | . $556{ }^{* *}$ | . $575{ }^{* *}$ | . $592{ }^{* *}$ | . $487{ }^{* *}$ | . $477{ }^{* *}$ | .549** | . $582{ }^{* *}$ | . $501{ }^{* *}$ | .583** | . $590{ }^{* *}$ |
| TERCT | . $518^{* *}$ | . $492{ }^{* *}$ | . $624^{* *}$ | . $538{ }^{* *}$ | . $621^{* *}$ | . $636{ }^{* *}$ | . $696{ }^{* *}$ | . $673 * *$ | .586** | .639** |
| CONLI | . $492{ }^{* *}$ | . $506{ }^{* *}$ | .585** | .591** | . $587{ }^{* *}$ | .520** | . $681{ }^{* *}$ | .658** | . $541{ }^{* *}$ | . $607^{* *}$ |
| COMLI | . $391{ }^{* *}$ | . $412{ }^{* *}$ | . $386{ }^{* *}$ | . $525^{* *}$ | . 363 ** | . $483{ }^{* *}$ | .598** | . $461{ }^{* *}$ | . $488{ }^{* *}$ | . $476{ }^{* *}$ |
| NAPDF | . $432{ }^{* *}$ | . $426{ }^{* *}$ | . $429^{* *}$ | . $465^{* *}$ | . $461{ }^{* *}$ | . $497{ }^{* *}$ | . $505{ }^{* *}$ | . 423 ** | . $426{ }^{* *}$ | . $396{ }^{* *}$ |
| LFDF | . $408^{* *}$ | . $439^{* *}$ | . $437{ }^{* *}$ | .508** | . $497{ }^{* *}$ | . $475^{* *}$ | . $529^{* *}$ | .504** | . $500^{* *}$ | . $408^{* *}$ |
| RFDF | . 463 ** | . $495{ }^{* *}$ | . $547{ }^{* *}$ | . $429^{* *}$ | . 523 ** | . $541{ }^{* *}$ | . $557{ }^{* *}$ | . $492{ }^{* *}$ | . $490{ }^{* *}$ | . $521^{* *}$ |
| IFDF | . $491{ }^{* *}$ | . $484^{* *}$ | . $505^{* *}$ | . $372{ }^{* *}$ | . $447{ }^{* *}$ | . $502{ }^{* *}$ | . $488{ }^{* *}$ | .403** | . $401{ }^{* *}$ | . $454{ }^{* *}$ |
| IMDF | . $421^{* *}$ | . $452{ }^{* *}$ | . $342{ }^{* *}$ | . $335^{* *}$ | . $418{ }^{* *}$ | . $447{ }^{* *}$ | . $406{ }^{* *}$ | . $308{ }^{* *}$ | . $456{ }^{* *}$ | . $348{ }^{* *}$ |
| PADF | . $437^{* *}$ | . $441^{* *}$ | . $356{ }^{* *}$ | . $345{ }^{* *}$ | . $450{ }^{* *}$ | . $408{ }^{* *}$ | . $389^{* *}$ | . $308{ }^{* *}$ | . $372{ }^{* *}$ | . $411{ }^{* *}$ |
| POLWI | . $386{ }^{* *}$ | . 391 ** | . $344{ }^{* *}$ | . 353 ** | . $474 * *$ | . $428{ }^{* *}$ | .548** | .419** | . $418{ }^{* *}$ | . $405^{* *}$ |
| ICDF | . $396{ }^{* *}$ | . 413 ** | . $463{ }^{* *}$ | . $412{ }^{* *}$ | . 379 ** | . $416{ }^{* *}$ | . $449^{* *}$ | . 380 ** | . $356{ }^{* *}$ | . $346{ }^{* *}$ |
| FRDR | . 373 ** | . 353 ** | . $418^{* *}$ | . $346{ }^{* *}$ | . 390 ** | . $412^{* *}$ | . $410{ }^{* *}$ | . $369^{* *}$ | . $345{ }^{* *}$ | . $284{ }^{* *}$ |
| DFPAPM | . $437^{* *}$ | . $432{ }^{* *}$ | . $395{ }^{* *}$ | . $345{ }^{* *}$ | . $386{ }^{* *}$ | . $411{ }^{* *}$ | . $433{ }^{* *}$ | . 370 ** | . $405^{* *}$ | . $339^{* *}$ |
| DFRACM | . $410{ }^{* *}$ | . 372 ** | . $347{ }^{* *}$ | . $384^{* *}$ | . 406 ** | . $400^{* *}$ | . $429^{* *}$ | . $335^{* *}$ | . $384{ }^{* *}$ | . 350 ** |
| PPAC | . $388{ }^{* *}$ | . $404^{* *}$ | . $365^{* *}$ | . 321 ** | . 387 ** | . $410{ }^{* *}$ | . $368{ }^{* *}$ | . 351 ** | . $408^{* *}$ | . $312{ }^{* *}$ |
| ADFT | . 391 ** | . $492{ }^{* *}$ | . $329^{* *}$ | . 357 ** | . $404{ }^{* *}$ | . $354{ }^{* *}$ | . $395{ }^{* *}$ | . 326 ** | . $437{ }^{* *}$ | . $343{ }^{* *}$ |
| RTDFP | . $354{ }^{* *}$ | . 420 ** | . $375^{* *}$ | . $292{ }^{* *}$ | .285** | . $320{ }^{* *}$ | . $330{ }^{* *}$ | . $234 * *$ | . $325^{* *}$ | . $265{ }^{* *}$ |
| ECSM | . 343 ** | . 361 ** | . $316{ }^{* *}$ | . $339^{* *}$ | . $345{ }^{* *}$ | . $425^{* *}$ | . 409 ** | . 310 ** | . $342{ }^{* *}$ | . $327{ }^{* *}$ |
| TLRR | . 249 ** | . $248{ }^{* *}$ | . $182^{* *}$ | . 231 ** | .154* | . $188{ }^{* *}$ | .204** | .175* | .199** | .142* |
| TLIN | . 431 ** | . $405^{* *}$ | . $389^{* *}$ | . $492{ }^{* *}$ | . $346{ }^{* *}$ | . $400^{* *}$ | . $365{ }^{* *}$ | . $429^{* *}$ | . $343{ }^{* *}$ | . $325^{* *}$ |
| TLPE | . $526^{* *}$ | . $441{ }^{* *}$ | . $516^{* *}$ | . 433 ** | . $411{ }^{* *}$ | . $443{ }^{* *}$ | . 452 ** | . $426{ }^{* *}$ | . $346{ }^{* *}$ | . $357{ }^{* *}$ |
| DBDFAI | . $500^{* *}$ | . 433 ** | . $497{ }^{* *}$ | . $454{ }^{* *}$ | . $451{ }^{* *}$ | . $474^{* *}$ | . $470{ }^{* *}$ | . $389^{* *}$ | . $392{ }^{* *}$ | . $466{ }^{* *}$ |


| Correlations Matrix |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | HACK | SOENG | TERCT | CONLI | COMLI | NAPDF | LFDF | RFDF |
| DEIP | . 626 | . 556 | . 518 | . 492 | . 391 | . 432 | . 408 | . 463 |
| DECV | . 591 | . 575 | . 492 | . 506 | . 412 | . 426 | . 439 | . 495 |
| SDDSM | . 643 | . 592 | . 624 | . 585 | . 386 | . 429 | . 437 | . 547 |
| RAPSM | . 672 | . 487 | . 538 | . 591 | . 525 | . 465 | . 508 | . 429 |
| ADCDD | . 671 | . 477 | . 621 | . 587 | . 363 | . 461 | . 497 | . 523 |
| INTH | . 668 | . 549 | . 636 | . 520 | . 483 | . 497 | . 475 | . 541 |
| EIAM | . 739 | . 582 | . 696 | . 681 | . 598 | . 505 | . 529 | . 557 |
| INBIN | . 677 | . 501 | . 673 | . 658 | . 461 | . 423 | . 504 | . 492 |
| ONFR | . 704 | . 583 | . 586 | . 541 | . 488 | . 426 | . 500 | . 490 |
| INES | . 587 | . 590 | . 639 | . 607 | . 476 | . 396 | . 408 | . 521 |
| HACK | 1 | . 603 | . 646 | . 582 | . 505 | . 580 | . 575 | . 584 |
| SOENG | . 603 | 1 | . 645 | . 446 | . 452 | . 477 | . 449 | . 502 |
| TERCT | . 646 | . 645 | 1 | . 628 | . 523 | . 467 | . 467 | . 530 |
| CONLI | . 582 | . 446 | . 628 | 1 | . 598 | . 423 | . 505 | . 514 |
| COMLI | . 505 | . 452 | . 523 | . 598 | 1 | . 375 | . 457 | . 417 |
| NAPDF | . 580 | . 477 | . 467 | . 423 | . 375 | 1 | . 791 | . 773 |
| LFDF | . 575 | .449 | .467 | . 505 | . 457 | .791 | 1 | . 782 |
| RFDF | . 584 | . 502 | . 530 | . 514 | . 417 | . 773 | . 782 | 1 |
| IFDF | . 513 | . 476 | . 531 | . 452 | . 392 | . 652 | . 687 | . 764 |
| IMDF | . 460 | . 465 | . 388 | . 333 | . 316 | . 648 | . 598 | . 638 |
| PADF | . 457 | . 434 | . 396 | . 367 | . 219 | . 564 | . 538 | . 556 |
| POLWI | . 495 | . 362 | .413 | . 476 | . 435 | . 530 | . 520 | . 632 |
| ICDF | . 499 | . 414 | . 398 | .408 | . 418 | . 605 | . 695 | . 609 |
| FRDR | . 531 | . 420 | . 401 | . 313 | . 329 | . 684 | . 651 | . 628 |
| DFPAPM | . 480 | . 402 | .418 | . 365 | . 321 | . 711 | . 658 | . 660 |
| DFRACM | . 484 | . 438 | . 386 | . 350 | . 351 | . 692 | . 653 | . 563 |
| PPAC | . 453 | . 425 | . 362 | . 315 | . 343 | . 588 | . 612 | . 620 |
| ADFT | . 449 | . 430 | . 326 | . 310 | . 323 | . 549 | . 616 | . 610 |
| RTDFP | . 383 | .463 | . 296 | . 239 | . 252 | . 525 | . 586 | . 540 |
| ECSM | . 442 | . 412 | . 404 | . 357 | . 336 | . 627 | . 635 | . 615 |
| TLRR | . 237 | . 210 | . 148 | . 187 | . 224 | . 329 | . 325 | . 288 |
| TLIN | . 444 | . 416 | . 390 | . 386 | . 423 | . 497 | . 588 | . 480 |
| TLPE | . 542 | . 408 | . 475 | . 471 | . 355 | . 520 | . 523 | . 505 |
| DBDFAI | . 568 | . 541 | . 432 | .435 | . 286 | . 579 | . 508 | . 543 |


| Correlations |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | IFDF | IMDF | PADF | POLWI | ICDF | FRDR | DFPAPM | DFRACM |
| DEIP | . 491 | . 421 | . 437 | . 386 | . 396 | . 373 | . 437 | . 410 |
| DECV | . 484 | . 452 | . 441 | . 391 | . 413 | . 353 | . 432 | . 372 |
| SDDSM | . 505 | . 342 | . 356 | . 344 | . 463 | . 418 | . 395 | . 347 |
| RAPSM | . 372 | . 335 | . 345 | . 353 | . 412 | . 346 | . 345 | . 384 |
| ADCDD | . 447 | . 418 | . 450 | . 474 | . 379 | . 390 | . 386 | . 406 |
| INTH | . 502 | . 447 | . 408 | . 428 | . 416 | . 412 | . 411 | . 400 |
| EIAM | . 488 | . 406 | . 389 | . 548 | . 449 | . 410 | . 433 | . 429 |
| INBIN | . 403 | . 308 | . 308 | . 419 | . 380 | . 369 | . 370 | . 335 |
| ONFR | . 401 | . 456 | . 372 | . 418 | . 356 | . 345 | . 405 | . 384 |
| INES | . 454 | . 348 | . 411 | . 405 | . 346 | . 284 | . 339 | . 350 |
| HACK | . 513 | . 460 | . 457 | . 495 | . 499 | . 531 | . 480 | . 484 |
| SOENG | . 476 | . 465 | . 434 | . 362 | . 414 | . 420 | . 402 | . 438 |
| TERCT | . 531 | . 388 | . 396 | . 413 | . 398 | . 401 | . 418 | . 386 |
| CONLI | . 452 | . 333 | . 367 | . 476 | . 408 | . 313 | . 365 | . 350 |
| COMLI | . 392 | . 316 | . 219 | . 435 | . 418 | . 329 | . 321 | . 351 |
| NAPDF | . 652 | . 648 | . 564 | . 530 | . 605 | . 684 | . 711 | . 692 |
| LFDF | . 687 | . 598 | . 538 | . 520 | . 695 | . 651 | . 658 | . 653 |
| RFDF | . 764 | . 638 | . 556 | . 632 | . 609 | . 628 | . 660 | . 563 |
| IFDF | 1 | . 712 | . 674 | . 674 | . 722 | . 614 | . 671 | . 625 |
| IMDF | . 712 | 1 | . 786 | . 695 | . 592 | . 654 | . 718 | . 726 |
| PADF | . 674 | . 786 | 1 | . 683 | . 676 | . 653 | . 723 | . 737 |
| POLWI | . 674 | . 695 | . 683 | 1 | . 631 | . 637 | . 689 | . 598 |
| ICDF | . 722 | . 592 | . 676 | . 631 | 1 | . 702 | . 702 | . 693 |
| FRDR | . 614 | . 654 | . 653 | . 637 | . 702 | 1 | . 770 | . 738 |
| DFPAPM | . 671 | . 718 | . 723 | . 689 | . 702 | . 770 | 1 | . 790 |
| DFRACM | . 625 | . 726 | . 737 | . 598 | . 693 | . 738 | . 790 | 1 |
| PPAC | . 711 | . 670 | . 672 | . 668 | . 694 | . 677 | . 757 | . 726 |
| ADFT | . 692 | . 644 | . 637 | . 627 | . 644 | . 638 | . 669 | . 690 |
| RTDFP | . 656 | . 591 | . 560 | . 538 | . 672 | . 651 | . 631 | . 610 |
| ECSM | . 610 | . 598 | . 536 | . 562 | . 573 | . 692 | . 677 | . 592 |
| TLRR | . 281 | . 252 | . 219 | . 265 | . 299 | . 320 | . 322 | . 277 |
| TLIN | . 518 | . 420 | . 425 | . 460 | . 627 | . 565 | . 557 | . 517 |
| TLPE | . 622 | . 549 | . 519 | . 576 | . 637 | . 601 | . 589 | . 605 |
| DBDFAI | . 509 | . 560 | . 559 | . 492 | . 457 | . 541 | . 557 | . 580 |


| Correlations |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | PPAC | ADFT | RTDFP | ECSM | TLRR | TLIN | TLPE | DBDFAI |
| DEIP | . 388 | . 391 | . 354 | . 343 | . 249 | .431 | . 526 | . 500 |
| DECV | . 404 | . 492 | . 420 | . 361 | . 248 | . 405 | . 441 | . 433 |
| SDDSM | . 365 | . 329 | . 375 | . 316 | . 182 | . 389 | . 516 | . 497 |
| RAPSM | . 321 | . 357 | . 292 | . 339 | . 231 | . 492 | . 433 | 454 |
| ADCDD | . 387 | . 404 | . 285 | . 345 | . 154 | . 346 | 411 | 451 |
| INTH | . 410 | . 354 | . 320 | . 425 | . 188 | . 400 | . 443 | . 474 |
| EIAM | . 368 | . 395 | . 330 | . 409 | . 204 | . 365 | . 452 | . 470 |
| INBIN | . 351 | . 326 | . 234 | . 310 | . 175 | 429 | . 426 | . 389 |
| ONFR | . 408 | . 437 | . 325 | . 342 | . 199 | . 343 | . 346 | . 392 |
| INES | . 312 | . 343 | . 265 | . 327 | . 142 | . 325 | . 357 | 466 |
| HACK | 453 | . 449 | . 383 | . 442 | . 237 | . 444 | . 542 | . 568 |
| SOENG | . 425 | . 430 | . 463 | .412 | . 210 | . 416 | . 408 | . 541 |
| TERCT | . 362 | . 326 | . 296 | . 404 | . 148 | . 390 | . 475 | . 432 |
| CONLI | . 315 | . 310 | . 239 | . 357 | . 187 | . 386 | . 471 | . 435 |
| COMLI | . 343 | . 323 | . 252 | . 336 | . 224 | . 423 | . 355 | . 286 |
| NAPDF | . 588 | . 549 | . 525 | . 627 | .329 | 497 | . 520 | . 579 |
| LFDF | . 612 | . 616 | . 586 | . 635 | . 325 | . 588 | . 523 | . 508 |
| RFDF | . 620 | . 610 | . 540 | . 615 | . 288 | . 480 | . 505 | . 543 |
| IFDF | . 711 | . 692 | . 656 | . 610 | . 281 | . 518 | . 622 | . 509 |
| IMDF | . 670 | . 644 | . 591 | . 598 | .252 | . 420 | . 549 | . 560 |
| PADF | . 672 | . 637 | . 560 | . 536 | . 219 | . 425 | . 519 | . 559 |
| POLWI | . 668 | . 627 | . 538 | . 562 | . 265 | 460 | . 576 | . 492 |
| ICDF | . 694 | . 644 | . 672 | . 573 | . 299 | . 627 | . 637 | . 457 |
| FRDR | . 677 | . 638 | . 651 | . 692 | . 320 | . 565 | . 601 | . 541 |
| DFPAPM | . 757 | . 669 | . 631 | . 677 | . 322 | . 557 | . 589 | . 557 |
| DFRACM | . 726 | . 690 | . 610 | . 592 | . 277 | . 517 | . 605 | . 580 |
| PPAC | 1 | . 782 | . 696 | . 655 | . 306 | . 619 | . 612 | . 592 |
| ADFT | .782 | 1 | . 770 | . 661 | . 308 | . 568 | . 578 | . 478 |
| RTDFP | . 696 | . 770 | 1 | . 742 | . 341 | . 678 | . 659 | . 476 |
| ECSM | . 655 | . 661 | . 742 | 1 | . 376 | . 657 | . 591 | . 550 |
| TLRR | . 306 | . 308 | . 341 | . 376 | 1 | . 414 | . 362 | . 279 |
| TLIN | . 619 | . 568 | . 678 | . 657 | .414 | 1 | . 744 | . 560 |
| TLPE | . 612 | . 578 | . 659 | . 591 | . 362 | . 744 | 1 | . 680 |
| DBDFAI | . 592 | . 478 | . 476 | . 550 | . 279 | . 560 | . 680 | 1 |

