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Factor Analysis-Based Investigation into Financial Crime Related Issues in Nigeria

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I. INTRODUCTION

inancial crime is a non-violent but intentional crime committed for illicit monetary or other unlawful gain from individuals, corporations, government bodies and financial institution (IMF, 2001; Ladan, 2005). It constitutes a very serious threat that manifests itself in virtually all aspects of national life. It is widely spread and involves Internet-based cheque issuance, cash deposit, wire transfer and Automated Teller Machine (ATM) withdrawals performed by institutions, government and individuals on a daily basis. Notable financial crimes include theft, scams, embezzlement, identity theft, money laundering, bribery, smuggling, forgery, counterfeiting and tax evasion (Ibrahim et al., 2015). Financial crimes are characterized by deceit. concealment or violation of trust and can be committed with every form of dynamism, subtleness and camouflage (Agus et al., 2010a). Considerably, financial crime may lead to colossal loss of money as well as undermining the integrity of financial institutions and markets. It may also subvert national growth and

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development (Spencher and Pickett, 2002; McDowell and Novis, 2001; Okoye and Gbegi, 2013; Ejiofor *et al.*, 2007). Financial crimes may lessen the ability of a country to attract foreign investment and subvert the growth and development of local manufacturing industries (IMF, 2001; Spencher and Pickett, 2002; Yusu, 2009). Financial crime may manifest inform of corruption, fraud and theft. Corruption is any illegal act such as kickbacks, embezzlement and extortion and another misuse of entrusted funds and power for private gain or improper and unlawful enrichment (Gottschalk, 2010; Ksenia, 2008).

Fraud is a despicable act with the aim of achieving a personal gain or creating a loss for another through concealment of an illegal act and it is a significant and growing threat to several organizations (Golden et al., 2006; Edelherz, 1977). Most prominent financial frauds are the conversion of public money into personal use, granting of unauthorized loans or overdraft, fraudulent transfer or withdrawals, misrepresentation of quality and quantity during procurement and pyramid trading schemes. Others are posting of fictitious credits, cheque counterfeiting or forging, payroll padding (ghost workers), contract over billings and over-invoicing among others (Okeyi and Gbegi, 2013). Theft of cash, intellect, art or identity is said to take place if it is carried out unlawfully and out of all proportions. Several strategies and measures for combating financial crimes include the use of technology and establishment of agencies and commissions. Technological tools offer a more holistic view of data and highlight potential areas of risk to organizations thereby reducing the incidence of fraud (Deloitte, 2014). Big data and text mining, machine learning and forensic accounting are some of the existing technologies for combating financial crimes (Adegbeie and Fakile, 2012; Shai and Shai, 2014; Agus et al., 2010b; Raghavendra et al., 2011; Kitten, 2016). Impact of criminal personality, opportunity structures, corporate identity, values on crime, and business ethics had been identified as causes of financial crimes (Bussman, 2003). These causes could be attributed to bio-genetic factors which include genetic mutation and hereditv (Horton. 1939). psychological factors comprising of personality disorders and sociological factors that include learning environment (Sutherland, 1939).

The fundamental techniques for combating financial crime still require a good understanding of its causes and dynamics as all technical and scientific proof have yielded unsatisfactory results (Ayoola eat al., 2015). Existing techniques for presenting the understanding of the causes and dynamics of financial crimes include some baseline and dimension-reduction tools such as Missing Values Ratio (MVR), Low Variance Filter (LVF) and High Correlation Filter (HCF). Others are Random Forests Ensemble Trees (RFET), Backward Feature Elimination (BFE), Forward Feature Construction (FFC), Principal Component Analysis (PCA) (Silipo, 2015) and Factor Analysis (FA). Factor analysis is a method for investigating whether some variables of interest N_1 , N_2 , . . ., N_m , are linearly related to a smaller number of unobservable factors F_1 , F_2 , . . ., F_k . It is used to identify dimensions underlying response (outcome) for a set of variables such that the observed values for the variables manifest are determined as variables. Some standardized variables are generally used with the correlation matrix modelled such that its dimensions correspond to the factors. Several of manifest variables can be used but more appropriate if they have more than a few distinct values and an approximate bellshaped distribution. Factor analysis based on principal components uses weights and scores to produce factor loadings and scores. These attributes informed its choice for the analysis of the financial security-related issues in Nigeria. The main objective of the study is to take a holistic view of the conceptualization of the main issues that relate to financial crime and provides data that serve the basis for the determination of their impact in Nigeria. Also, the study also provides data that is relevant for drawing conclusion based on a comparison between results from current and some related works.

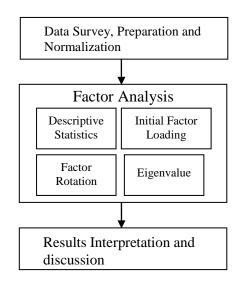
II. Related Works

An exploration of the statistical methods for fighting financial crime by financial institutions is carried out in (Agus et al., 2010a). Issues on the growing losses of revenue by governments, financial institution and individuals to the various forms of financial crime as well as the review of some statistical techniques for investigative studies of financial crimes were also discussed. The research formulated the necessary account opening, described steps for some visualization, description, analysis and computational tools for data on high volume transactions as well as a machine learning algorithm for detecting financial crime. An investigative study on the impact of economic and financial crime on the Nigerian economy is presented in (Yusus, 2009). A review of the conceptual legal framework as well as the nature, scope and effects of economic and financial crimes under the Nigerian law, was presented. The authors concluded that ICT infrastructure is the main tool that financial criminals rely on in carrying out their unlawful acts.

The authors in (Okoye and Gbegi, 2013) evaluated the effect of fraud and related financial crime on the Nigerian economy. The research placed a premium on how the Internet, electronic money transfer (wire transfer) and other related platforms contributed to the current spate of financial crime. Regression-based analysis on available financial crime data revealed that financial crimes portend dwindling Gross Domestic Product (GDP) and shrinking economy. In (Iwasokun et al., 2012), an investigation on the effect of financial crime on the society was presented. A platform for determining the correlation between crimes was also presented based on PCA-based analysis of financial crime data from a Criminal Investigation Department. The authors in [3] examined the effect of financial crime corruption on manufacturing and firms and organizations. The Two Way-ANOVA-based analysis of financial crime data obtained from primary and secondary sources revealed significant and negative implications of financial crime on the manufacturing firms as manifested through drained revenue, operational instability and low level of interest from foreign and domestic investors.

III. Research Methodology

The research methodology is conceptualized in Figure 1 showing data survey, factor analysis by principal components and results phases.



(3)

statistics ensured standardization of the measurements used in the normalized data and covered the assigning of the data set variables to zero means, unit variances and standard deviation. The sample correlation coefficient is calculated as follows: $r_{x,y} = \frac{s_{x,y}}{s_x s_y}$

 $s_{x,y}$ is the variance between two columns in the data matrix, s_x and s_y are the standard deviations of columns x and y respectively. For multivariate analysis, the correlation matrix is analogous to the covariance matrix with correlations in place of covariances. Barlett's test of sphericity β is used to confirm the adequacy of a sample population by testing the null hypothesis that the variables in the population correlation matrix are uncorrelated. The observed significance level of .0000 is used to reject this hypothesis. The test is based on the formula (Donal, 1993):

$$\beta = -\left[[n - 1 - \frac{2p + 5}{6}] \right] ln|R|$$
 (4)

|R| is the determinant of the correlation matrix, n is the number of observation and p is the number of variables. The KMO test γ for the hypothesis that the variables are uncorrelated in the population is based on the formula:

$$\gamma = \frac{\sum_{i} \sum_{i \neq j} r_{ij}^{2}}{\sum_{i} \sum_{i \neq j} r_{ij}^{2} + \sum_{i} \sum_{i \neq j} a_{ij}^{2}}$$
(5)

 r_{ij} is the correlation coefficient in the correlation matrix; a_{ii} is the partial correlation coefficient and i,j represent the rows and column sizes respectively. A near-zero partial correlation, A guarantees effective factorization and it is obtained from the correlation matrix R as follows:

Figure 1: Conceptualization of the Factor Analysis-based model

Data Survey, preparation and Normalization a)

Data survey involved a survey of public and private agencies such as banks, insurance company, educational institutions that are involved in different forms of financial activities both offline and online. The selection was based on stratified sampling and respondents were randomly chosen with equal probability. Data preparation involved determination of all relevant variables for inclusion in the analysis, determination of the number of observations sufficient to provide reliable estimations of the correlations between the indices, estimation of valid measure of associations among selected variables and the arrangement of the surveyed observations as a set of data vectors y_1, y_2, \dots, y_d with each denoting a grouped observation of V variables. Data normalization is used to transform the surveyed data to a formatted form. Data with ratings were restructured to a notionally common scale before averaging.

b) PCA-Based Factor Analysis

The variables for PCA-based factor analysis of the inducement factors of financial crime are related to one another for the jth respondents, and it is represented as follows (Iwasokun et al., 2012; Gulumbe et al., 2012):

$$var(Z_j) = \lambda_j \tag{1}$$

$$\alpha_{j1}^{2} + \alpha_{j2}^{2} + \dots + \alpha_{jk}^{2} = 1$$
 (2)

Z_i represents the jth respondent (the principal component of j^{th} data), α_{ik} represents the assessment of the j^{th} indices by the kth respondent (the elements of the jth eigenvector λ_i for the correlation matrix).

The principal components analysis of the survey involved descriptive statistic, correlation matrix, Bartlett's and Kaiser-Mayer Olkin (KMO) tests, component extraction and other statistics of relevance. Descriptive

$$A = -\frac{1}{R \times \sqrt{v_{ii} \times v_{jj}}}$$
(6)

The Communality of the squared component loadings for component *i* is computed as follows:

$$c_i = \alpha_{i1}^2 + \alpha_{i2}^2 + \dots + \alpha_{ip}^2 = \sum_{i=1}^p \alpha_{ip}^2$$
 (7)

p is the number of variables, α_{ip} is the value in A for row *i*, column *p*. The communalities narrate how well the model performs for each variable while the total communality gives an overall assessment.

The eigenvalues of R is calculated as follows:

$$\left|\mathbf{R} - \lambda \mathbf{I}_{\mathbf{p}}\right| = 0 \tag{8}$$

 I_p is a $p \times p$ identity matrix with eigenvalues $\widehat{\lambda_1} \ge \widehat{\lambda_2} \ge \cdots \ge \widehat{\lambda_p}$ and the eigenvector V is calculated as follows:

$$V = DR^{-1} \tag{9}$$

D is the $p \ge p$ diagonal matrix of eigenvalues of R.

From p variables, the principal components (unrotated factors) are extracted based on the criterion presented as follows (Kaiser, 1960):

$$\bar{\lambda} = \frac{1}{p} \sum_{j=1}^{p}$$
(10)

The criterion only extracted a principal component with an eigenvalue greater than $\bar{\lambda}$. The unrotated factors are subjected to orthogonal transformation using varimax, equimax, quartimax and promax and the best result was taken. Orthogonal transformation is used to obtain meaningful representation of variables and component mapping along the principal axis. Rotation by varimax is based on the assumption that the interpretability of a factor can be measured by the variance of the squares of its factor loadings. Quartimax rotation involves the minimization

of the number of factors needed to explain each variable while equamax rotation is a compromise that attempts to simplify both components and variables. Promax is an oblique rotation that allows factors to be correlated and because it is often more easily calculated than any direct oblimin rotation, it is more useful for large datasets.

The determination of component scores is based on a linear equation of the weighted standard scores of each respondent on the variables as follows:

$$M_{b,c} = \sum_{m=1}^{7} d_{a,c} W_{b,a}, b = 1, 2, ..., x; \quad m = 1, 2, ..., f$$
(11)

 $M_{b,c}$ represents the contribution of bth respondent to cth component, $d_{a,c}$ is the component score coefficient of ath variable for cth component, *f* is the number of the extracted components, $W_{b,a}$ represents the standard score of bth respondent for ath variable and x is the respondents' population. $W_{b,a}$ is estimated as follows:

$$W_{b,a} = X + \frac{S_b - T_b}{u_b}$$
 (12)

X represents the allowable minimum score, which in this case is 1, S_b represents the raw score for b^{th} index, T_b and u_b represent the mean and standard deviation respectively, of the raw scores of b^{th} index by the sampled respondents.

IV. Results and Interpretation

The result from the analysis is interpreted to determine the correlation and relationship between indices. The Questionnaire shown in Appendix 1 was designed using the indices for the analysis of financial security-related issues. Each of these indices was offered loosed linguistic description and range of values as shown in Table 1.

Table 1: Matri	k of Weight Attached	d to Linguistic Value

Linguistic Representation	Excellent	Very Good	Good	Average	Poor
Range of Values	4.01-5.0	3.01-4.0	2.01-3.0	1.01-2.0	0.0-1.0

The first part of the Questionnaire provided vital information about each respondent while the second part presented five columns for the respondent to rank each of the sixteen indices based on the scale presented in Table 1. The Questionnaire was administered to Thirty States in the six geo-political zones and the Federal Capital Territory (FCT) in Nigeria and the summary of the survey is presented in Table 2.

Serial No.	State	No. of LG Surveyed	Total Questionnaire Administered	Total Questionnaire Returned	Total Questionnaire not Returned
1	Abia	5	300	263	37
2	Adamawa	6	425	415	10
3	Akwa-Ibom	8	524	522	2
4	Anambra	5	275	254	21
5	Bauchi	7	589	487	102
6	Benue	7	652	623	29
7	Delta	10	524	451	73
8	Cross River	11	687	671	16
9	Ebonyi	6	165	128	37
10	Edo	8	785	687	98
11	Ekiti	7	570	457	113
12	Enugu	8	622	528	94
13	Imo	7	522	420	102
14	Jigawa	3	420	259	161
15	Kaduna	3	186	181	5
16	Kano	11	894	856	38
17	Kebbi	3	202	202	0
18	Kogi	6	414	401	13
19	Kwara	4	551	510	41
20	Lagos	20	1026	896	130
21	Nasarawa	6	239	239	0
22	Ogun	7	658	452	206
23	Niger	8	659	659	0
24	Ondo	18	1524	1325	199
25	Osun	8	354	258	96
26	Oyo	12	742	468	274
27	Plateau	6	231	197	34
28	Rivers	6	402	401	1
29	Sokoto	3	189	175	14
30	Taraba	5	580	574	6
31	FCT	3	627	587	40
	Total	158	11289	9500	1789

Table 2: Summary of the survey

A total of Sixteen Thousand Five Hundred and Thirty-Eight (16538) copies of the Questionnaire were administered through direct and online contacts. In the direct contact, the researcher visited the surveyed states or engaged the services of third parties while indirect contact involved hosting the Questionnaire on Google forms for online assessment. In both cases, Fourteen Thousand Five Hundred and Forty-Six (14546) respondents returned duly completed Questionnaires. Where necessary, the responses were verified and

validated through follow-up meetings and personal interviews with the respondents

All the 14546 responses were subjected to factor analysis by principal components using SPSS. The analysis of the respondents' knowledge of financial crime, times fallen victim of financial crime and the distribution of crimes are presented in Tables 3, 4 and 5 respectively.

Values	Frequency	Percent	% Cumulative
Poor	988	6.8	6.8
Average	3102	21.3	28.1
Good	4803	33.0	61.5
Very Good	3863	26.3	87.7
Excellent	1790	12.3	100.0
Total	14546	100.0	

Table 3: Knowledge of Financial Crime

Range	Frequency	equency Percent		Cumulative Percent
0	11655	80.1		
1-5	1372	9.4	47.4	47.4
6-10	447	3.1	15.4	62.8
11-15	164	1.1	5.7	68.5
16-20	357	2.5	12.4	80.9
> 20	551	3.3	19.1	100.0
Total	14546	100.0		

Table 4: Distribution Range of Occurrences

Table 5: Classes of Financial Crime Victims Distribution

Classes of Financial Crime	Number of Occurrences	%
Advance fee fraud	1177	12.72
Forgery (Fake Cheque)	834	9.02
Money Theft Through ATM	850	9.18
Kickbacks and Extortion	730	7.89
Embezzlement	775	8.37
Corruption and Bribery	1014	10.95
Fraud	865	9.34
Money Laundering	536	5.8
Identity theft	462	4.99
Counterfeit Money	700	7.57
Financial Grooming	640	6.92
Insider Trading	372	4.03
Phishing	297	3.22
Total	9252	100

Variables	Ν	Mean	Std. Deviation
National Policy on Financial operations and Security	1341	3.47	1.245
Legislative, Regulatory and Institutional Framework on Financial operations	1350	3.25	1.172
Legislative, Regulatory and Institutional Framework on Financial Security	1347	3.27	1.184
Implementation of Conventional Security in Financial Institution	1338	3.21	1.177
Implementation of Financial Security Policy	1340	3.13	1.219
Financial Crime Case Assessment	1338	3.15	1.210
Prosecution of Financial Criminals	1344	3.05	1.223
Proficiency of litigators on Financial Crime Cases	1323	3.13	1.223
Public Awareness on Financial Security	1344	3.17	1.196
IT Literacy of Conventional Security Personnel	1346	3.13	1.150
Availability of IT Security Facility at Financial Centres	1259	3.05	1.213
Capacity Building/ IT Staff Development	1268	3.04	1.197
Rapid Response to Financial Emergency by Security Agencies	1346	3.11	1.270
Development and Usability of Financial Crime Database System	1355	3.07	1.267
Collaboration Between Financial Agencies	1341	3.09	1.221
Availability of Independent/Private Financial Security organization	1343	3.03	1.235

The descriptive statistics shown in Table 6 presents the means and standard deviation of the rating of the indices for the analysis of the financial crime related issues by the respondents. The mean and standard deviation of the rating on 'National Policy on Financial operations and Security' are 3.47(69.0%) and 1.245 respectively while the mean and standard deviation of the rating on 'Legislative, Regulatory and Institutional Framework on Financial operations' are 3.25(65.0%) and 1.172 respectively. These values reveal that on the average, the respondents agreed that the 'National Policy on Financial Operations and Security' and 'Legislative, Regulatory and Institutional Framework on Financial Operations' are strong financial crime related issues. The interpretation is based on the matrix of weight attached to the linguistic values presented in Table 1. Similarly, standard deviations represent the statistical measure of dispersion from the mean for the

response values for 'National Policy on Financial Operations and Security' and for 'Legislative, Regulatory and Institutional Framework on Financial Operations' respectively. The communalities of the indices for financial crime related issues are presented in Table 7 showing that communalities of the 'National Policy on Financial operations and Security' and 'Legislative, Regulatory and Institutional Framework on Financial operations' are 0.719 and 0.731 respectively. These imply that 71.9% of the variance in 'National Policy on Financial operations and Security' can be explained by the extracted factors while the remaining 28.1% is attributed to extraneous factors. Similarly, 73.1% of the variance in 'Legislative, Regulatory and Institutional Framework on Financial operations' can be explained by the extracted factors while the remaining 26.9% is credited to extraneous factors.

Variables	Initial	Extraction
National Policy on Financial operations and Security	1.000	0.719
Legislature's, Regulatory and Institutional Framework on Financial operations	1.000	0.731
Legislature's, Regulatory and Institutional Framework on Financial Security	1.000	0.718
Implementation of Conventional Security in Financial Institution	1.000	0.648
Implementation of Financial Security Policy	1.000	0.664
Public Awareness on Financial Crime/Security	1.000	0.698
Development and Usability of Financial Crime Database System	1.000	0.795
IT Literacy of Conventional Security Personnel	1.000	0.69
Capacity Building/ IT Staff Development	1.000	0.761
Collaboration Between Financial Agencies	1.000	0.756
Availability of Independent/Private Financial Security organization	1.000	0.68
Availability of IT Security Facility at Financial Centres	1.000	0.791
Proficiency of litigators on Financial Crime Cases	1.000	0.788
Financial Crime Case Assessment	1.000	0.623
Rapid Response to Financial Emergency by Security Agencies	1.000	0.621
Prosecution of Financial Criminals	1.000	0.681

Table 7: Communalities of Variables

Variables	NpFos	AsFrO	AsFrS	CsAss	FsAss	FCCcAs	ProFC	LigPr	PubAr	ltLit	FCSEc	CapSt	RapRe	FCDbs	Collb	InOrg
NpFos	1.000	.711	.656	.573	.542	.558	.473	.441	.421	.432	.383	.376	.451	.458	.470	.420
AsFrO	.711	1.000	.710	.594	.556	.570	.500	.460	.449	.493	.439	.441	.515	.493	.489	.445
AsFrS	.656	.710	1.000	.604	.565	.572	.522	.523	.466	.476	.466	.446	.484	.488	.478	.421
CsAss	.573	.594	.604	1.000	.632	.585	.496	.472	.464	.464	.462	.431	.500	.497	.518	.451
FsAss	.542	.556	.565	.632	1.000	.635	.442	.419	.431	.485	.448	.483	.489	.453	.489	.468
FCCcAs	.558	.570	.572	.585	.635	1.000	.565	.481	.455	.456	.471	.438	.471	.463	.481	.476
ProFC	.473	.500	.522	.496	.442	.565	1.000	.631	.502	.488	.457	.416	.512	.498	.491	.471
LigPr	.441	.460	.523	.472	.419	.481	.631	1.000	.605	.519	.440	.453	.481	.517	.499	.455
PubAr	.421	.449	.466	.464	.431	.455	.502	.605	1.000	.593	.482	.460	.458	.489	.474	.454
ItLit	.432	.493	.476	.464	.485	.456	.488	.519	.593	1.000	.612	.558	.515	.498	.506	.471
FCSEc	.383	.439	.466	.462	.448	.471	.457	.440	.482	.612	1.000	.668	.531	.451	.475	.522
CapSt	.376	.441	.446	.431	.483	.438	.416	.453	.460	.558	.668	1.000	.525	.471	.501	.492
RapRe	.451	.515	.484	.500	.489	.471	.512	.481	.458	.515	.531	.525	1.000	.643	.540	.527
FCDbs	.458	.493	.488	.497	.453	.463	.498	.517	.489	.498	.451	.471	.643	1.000	.696	.590
Collb	.470	.489	.478	.518	.489	.481	.491	.499	.474	.506	.475	.501	.540	.696	1.000	.626
InOrg	.420	.445	.421	.451	.468	.476	.471	.455	.454	.471	.522	.492	.527	.590	.626	1.000

Table 8: Correlation Matrix of Variables

The analysis of the correlation matrix presented in Table 8 shows that the highest correlation of 0.711 exists between 'National Policy on Financial Operations and Security' and 'Legislative, Regulatory and Institutional Framework on Financial Operations'. The next highest correlation of 0.710 exists between 'Legislative, Regulatory and Institutional Framework on Financial Security' and 'Legislative, Regulatory and Institutional Framework on Financial Operations'. The implication of the former is that 'National Policy on Financial Operations and Security' is most likely to share the same factor with 'Legislative, Regulatory and Institutional Framework on Financial Operations'. Similarly, in the later, 'Legislative, Regulatory and Institutional Framework on Financial Security' and 'Legislative, Regulatory and Institutional Framework on Financial operations' will likely share the same factor.

The Least correlation of 0.376 exists between 'Capacity Building/ IT Staff Development' and 'National Policy on Financial Operations and Security'. This means that 'Capacity Building/ IT Staff Development' and 'National Policy on Financial Operations and Security' are not likely to share the same factor. The Barlett's test of sphericity produces a χ^2 of 7493.525 with a significance level of 0.000 which indicates that the sample population is adequate while the Kaiser-Mayer Olkin (KMO) test produced a measure of 0.950, which further confirmed the adequacy of the sample population.

The result of Kaiser Criterion based initial component extractions is presented in Table 9. The orthogonal transformation of the initial component extractions by varimax, promax, equamax and quartimax were carried out and the result obtained from the rotation by varimax, which is presented in Table 10, appeared most realistic and meaningful for interpretation among all others. Table 10 reveals four factors with their corresponding loadings.

Variables		Com	oonent	
valiables	1	2	3	4
Legislative, Regulatory and Institutional Framework on Financial Security	0.762			
Legislature's, Regulatory and Institutional Framework on Financial operations	0.761			
Implementation of Conventional Security in Financial Institution	0.750			
Collaboration Between Financial Agencies	0.748			
Development and Usability of Financial Crime Database System	0.746		-0.421	
Financial Crime Case Assessment	0.744			
Rapid Response to Financial Emergency by Security Agencies	0.740			
Implementation of Financial Security Policy	0.732			
IT Literacy of Conventional Security Personnel	0.732			
Prosecution of Financial Criminals	0.724			
National Policy on Financial operations and Security	0.717	-0.450		
Proficiency of litigators on Financial Crime Cases	0.717			-0.469
Availability of IT Security Facility at Financial Centres	0.708			
Availability of Independent/Private Financial Security organization	0.708			
Public Awareness on Financial Crime/Security	0.699			
Capacity Building/ IT Staff Development	0.695			

Table 9: Extracted Initial Component Matrix

Principal Component 1(Policies and Regulations) loads on National Policy on Financial Operations and Security, Legislature's, Regulatory and Institutional Framework on Financial Operations, Legislature's, Regulatory and Institutional Framework on Financial Security, Implementation of Financial Security Policy, Implementation of Conventional Security in Financial Institutions and Financial Crime Case Assessment. Principal Component 2 (Responses and Management) loads on Collaboration between Financial Agencies, Development and Usability of Financial Crime Database System, Availability of Independent/Private Financial Security organization and Rapid Response to Financial Emergency by Security Agencies.

Similarly, Principal Component 3 (Capacity Building) loads on Capacity Building/ IT Staff Development, Availability of IT Security Facility at Financial Centres and IT Literacy of Conventional Security Personnel. Principal Component 4 (Awareness and Litigation) loads on Proficiency of litigators on Financial Crime Cases, Public Awareness on Financial Crime and Security and Prosecution of Financial Criminals.

The result highlighted government-approved policies and regulation as the most critical issues on financial crimes. This view was corroborated by the authors in (Galina, 2014; Sofia de Oliveira et al., 2016) who mentioned that the state of national financial security depends solely on governance efficiency as well as policies and regulations. Other financial crimes related issues highlighted are response and management strategies, capacity building and public awareness and litigation measures. These also agreed with the opinions presented in (Galina, 2014; Sofia de Oliveira et al., 2016; Usman et al., 2012; Durmus, 2007). The Component Score which is a linear combination of the original variables to the extracted factors is presented in Table 11.

Variables	Component						
Variables	1	2	3	4			
National Policy on Financial operations and Security	0.788						
Legislature's, Regulatory and Institutional Framework on Financial operations	0.773						
Legislature's, Regulatory and Institutional Framework on Financial Security	0.744						
Implementation of Conventional Security in Financial Institution	0.688						
Implementation of Financial Security Policy	0.685						
Financial Crime Case Assessment	0.66						
Development and Usability of Financial Crime Database System		0.784					
Collaboration Between Financial Agencies		0.754					
Availability of Independent/Private Financial Security organization		0.701					
Rapid Response to Financial Emergency by Security Agencies		0.582					
Availability of IT Security Facility at Financial Centres			0.792				
Capacity Building/ IT Staff Development			0.77				
IT Literacy of Conventional Security Personnel			0.614	0.441			
Proficiency of litigators on Financial Crime Cases				0.783			
Public Awareness on Financial Crime and Security				0.691			
Prosecution of Financial Criminals				0.658			

Table 10: Rotated Component matrix

2019

Year

	Component								
Variables	1	2	3	4					
National Policy on Financial operations and Security	0.346	-0.063	-0.162	-0.06					
Legislative, Regulatory and Institutional Framework on Financial operations	0.316	-0.08	-0.077	-0.072					
Legislature's, Regulatory and Institutional Framework on Financial Security	0.288	-0.148	-0.066	0.033					
Implementation of Conventional Security in Financial Institution	0.252	-0.018	-0.032	-0.099					
Implementation of Financial Security Policy	0.270	-0.051	0.124	-0.246					
Public Awareness on Financial Crime/Security	-0.135	-0.159	0.061	0.493					
Development and Usability of Financial Crime Database System	-0.113	0.522	-0.209	-0.022					
IT Literacy of Conventional Security Personnel	-0.096	-0.166	0.35	0.167					
Capacity Building/ IT Staff Development	-0.081	-0.066	0.547	-0.171					
Collaboration Between Financial Agencies	-0.083	0.486	-0.138	-0.092					
Availability of Independent/Private Financial Security organization	-0.108	0.433	-0.001	-0.146					
Availability of IT Security Facility at Financial Centres	-0.083	-0.13	0.567	-0.112					
Proficiency of litigators on Financial Crime Cases	-0.125	-0.073	-0.156	0.602					
Financial Crime Case Assessment	0.228	-0.083	-0.012	-0.015					
Rapid Response to Financial Emergency by Security Agencies	-0.052	0.283	0.039	-0.087					
Prosecution of Financial Criminals	-0.031	-0.03	-0.185	0.454					

Table 12: Standard scores by Ten Respondents

Respond-nets	NpFos	AsFrO	AsFrS	CsAss	FsAss	FCcAs	ProFC	LigPr	PubAr	ltLit	FCSEc	CapSt	RapRe	FCDbs	Collb	InOrg
Res1	5	3	4	2	5	3	2	5	3	4	1	3	3	4	5	5
Res2	5	4	5	5	5	5	4	2	3	1	1	1	3	5	3	1
Res3	4	4	4	4	5	4	4	5	4	4	4	4	5	4	4	4
Res4	5	5	5	4	5	5	5	5	4	3	З	1	5	5	4	З
Res5	5	4	3	5	1	2	5	4	3	4	1	1	5	4	3	2
Res6	5	2	4	1	3	5	5	5	2	4	5	4	5	3	5	4
Res7	4	3	3	3	4	3	2	3	4	4	3	4	3	2	3	2
Res8	4	4	4	3	3	2	2	2	3	2	1	1	1	1	2	2
Res9	5	4	5	4	5	4	4	3	5	4	3	5	4	3	3	5
Res10	1	1	1	1	1	1	1	1	2	3	4	3	2	4	2	3

Given that the standard scores by the bth respondent in the sixteen variables under consideration are $W_{b,1}, W_{b,2}, W_{b,3} \ldots, W_{b,16}$, the financial crimes related issues based on the view of each respondent, in the areas of policies and regulations, responses and management, capacity building and awareness and litigation denoted by $M_1 M_{2}, M_{3}$, and M_4 are modeled as follows:

$$M_1 = 0.346W_{b,1} + 0.316W_{b,2} + \dots - 0.031W_{b,16}$$
(9)

$$M_2 = -0.063W_{b,1} - 0.800W_{b,2} + \dots - 0.30W_{b,16} \quad (10)$$

$$M_3 = -0.162W_{b,1} - 0.077W_{b,2} + \dots - 0.185W_{b,16} \quad (11)$$

$$M_4 = -0.060W_{b,1} - 0.072W_{b,2} + \dots + 0.454W_{b,16} \quad (12)$$

The standard scores by ten randomly selected respondents for each of the sixteen variables under consideration are presented in Table 12. Table 13 shows the calculated percentage contributions of each of the ten sampled respondents to each of the four factors. It is revealed that sampled respondent described with identity Res2 has the highest contribution of 5.76 (17.47%) to issue 1 while sampled respondent Res6 has the highest contribution of 3.89 (21.85%) to issue 2. Similarly, sampled respondent described with identity Res7 has the highest contribution of 3.29 (19.35%) to issue 3 and sampled respondent Res6 has the highest contribution of 4.50 (19.93%) to issue 4.

	Fa	ctor 1	Fa	ctor 2	Fa	ctor 3	Factor 4		
Respondents	Score	% Contribution	Score	% Contribution	Score	% Contribution	Score	% Contribution	
Res1	3.76	11.36	0.80	4.51	2.31	13.56	2.68	11.85	
Res2	5.79	17.47	0.02	0.10	0.65	3.84	1.62	7.19	
Res3	3.22	9.73	2.34	13.15	2.92	17.14	3.03	13.42	
Res4	4.62	13.96	1.83	10.28	1.08	6.32	3.51	15.54	
Res5	3.47	10.48	2.55	14.33	-0.30	-1.77	2.77	12.28	
Res6	1.17	3.52	3.89	21.85	1.57	9.20	4.50	19.93	
Res7	2.66	8.04	1.74	9.77	3.29	19.35	0.78	3.47	
Res8	4.02	12.12	0.45	2.53	0.93	5.47	0.21	0.93	
Res9	3.91	11.80	1.67	9.41	3.05	17.91	2.49	11.01	
Res10	0.50	1.51	2.50	14.08	1.53	8.98	0.99	4.38	
Total	33.12	100.00	17.78	100.00	17.03	100.00	22.58	100.00	

Table 13: Aggregate factor scores with percentage contributions for a subset of respondents

The eigenvalues and percentage variance for each of the four issues is shown in Table 14. It is revealed that the four extracted issues contributed 71.02% to financial crime related issues in Nigeria. Component 1 described as 'Policies and Regulations' contributes 53.37% out of 71.02%. This implies that government policies and regulations are very germane issues of financial security and must not be taken with levity. Components 2, 3 and 4 contribute 7.43%, 5.13% and 5.10% respectively. These imply that government must also focus on raising the awareness of its people

to the need for safe and secure financial system as well as ensuring strong litigation measures against financial related crimes. It is also important that adequate attention be given to building strong capacity for all relevant groups and agencies as well as putting in place facilities for timely response and management of threats to safe financial system. The remaining 28.98% is considered as the expected influences of some extraneous components that are important but their related indices were not considered.

Table 14: Eigenvalue of factors	S
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Component	Extra	ction Sums of Sq	uared Loadings
Component	Total	% of Variance	Cumulative %
1	8.539	53.367	53.367
2	1.188	7.426	60.793
3	.821	5.131	65.925
4	.815	5.095	71.020

V. CONCLUSION

Nigerian economy has suffered greatly in recent years dues to the rising cases of financial crimes in several agencies of public and private sectors. Financial criminals have used the Internet to commit all manners of frauds, embezzlement, tax invasion, money laundering and other despicable financial acts. Nigeria's international image has also suffered due to the involvement of some of her citizens in cases of financial crimes at local and international levels. In view of these, factor analysis by principal components has been used for the analysis of financial crimes related issues in Nigeria. Four issues were extracted with their respective related indices. The initial component matrix generated was subjected to orthogonal transformation to ensure reasonable factorization. The obtained factor score coefficient matrix provided the basis for the determination of the degree of reasonability of the assessment of every respondent.

The obtained eigenvalue of each issue gave its percentage impact on the current spate of financial

crimes in Nigeria. The percentage contribution was less than 100, which is a pointer to some significant extraneous (latent) factors whose indices were not considered in the research instrument. The results of the factor analysis placed a high premium on government policies and regulations, responses and management, capacity building as well as awareness and litigation as the major issues for safe and secure financial system in Nigeria. These corroborated the positions held by the authors in [30-33] who opined that good governance should be provided at all levels for economic and social security, promoting selflessness and patriotism. The authors also agreed on the need for adequate countermeasure and litigation systems as necessary strategies for curbing the menace of financial crime. Findings from the research also established that systemic ways of ensuring that citizens adopt technical know-how for national development rather than committing crimes should be introduced and enforced by the Nigerian aovernment.

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Appendix 1

Questionnaire on Analysis of Financial Crime Related Issues in the Scenery of Nigeria

The purpose of this Questionnaire is to conduct investigative analysis of financial crimes related issues in Nigerian with a view to developing a pro-active solution. Confidentiality of personal information is guaranteed. We would therefore appreciate your sincere contributions to the research by giving a very accurate and honest response to this Questionnaire.

Section A: Profile of Respondents

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10. *Have You Been a Victim of Computer Crime* (Please Tick ($\sqrt{}$) as appropriate)

Yes	
No	
IF YE	ES:

11. TICK $(\sqrt{)}$ AS APPROPRIATE, THE INCIDENCES

Advance fee fraud ("Yahoo Yahoo")	
Forgery (Fake Office Documents, Certificates, etc.)	
ATM (Money Theft Through ATM)	
Piracy (Pirated Software, Video/Audio CDs, etc.).	
Phreaking (Making Fraudulent free calls)	
Spamming (Unsolicited emails)	
Embezzlement (Executive Theft, Salami Shaving. etc.).	
Computer Virus and/or Denial of Service	
Pornography/Financial Grooming	
Others(Specify):	

12. TICK ($\sqrt{}$) AS MANY AS POSSIBLE MODE OF OCCURRENCES

1-5	
6-10	
11-15	
16-20	
Above 20	

13. Financial Crime Gender Incidence

Index	Very High	High	Medium	Low
Male				
Female				

14. Financial Criminals' Age Range

Index	Very High	High	Medium	Low
2-11				
12-17				
18-25				
26-45				
Above 45				

Section B: Assessment of Financial Security Related Issues

Pls. Tick/Write as appropriate depending on the level/Intensity of Indices using the scale of Excellent, Very Good, Good Average or Poor.

Index	Excellent	Very Good	Good	Average	Poor
National Policy on Financial operations and Security					
Legislative, Regulatory and Institutional Framework on					
Financial operations					
Legislative, Regulatory and Institutional Framework on					
Financial Security					
Implementation of Conventional Security in Financial					
Institution					
Implementation of Financial Security Policy					
Financial Crime Case Assessment					
Prosecution of Financial Criminals					
Proficiency of litigators on Financial Crime Cases					
Public Awareness on Financial Security					
IT Literacy of Conventional Security Personnel					
Availability of IT Security Facility at Financial Centers					
Capacity Building/ IT Staff Development					
Rapid Response to Financial Emergency by Security					
Agencies					
Development and Usability of Financial Crime Database					
System					
Collaboration Between Financial Agencies					
Availability of Independent/Private Financial Security					
organization					

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