Exploring the Factors Affecting the ICT Technology Transfer Process: An Empirical Study in Libya

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Abstract

The purpose of this work is to develop a model that describes the international technology transfer (ITT) process of information and communication technology (ICT) from developed to developing countries. This paper is a part of an ongoing study aiming to develop a technology transfer model examining the embracing of a foreign advanced technology to the ICT companies as well as ICT-based SME's projects in Libya. The past relevant TT models are reviewed with the intention of exploring and sort out the most ITT influential factors. The questionnaire that conducted recently in the TT process in the Libyan ICT industry was utilized to verify the model. Major statistical techniques are applied to analyze the survey received data. To establish reliable measures for the factors and sub-factors under investigation as well as to reduce their numbers, Exploratory Factor Analysis (EFA) was implemented. Furthermore, the goal behind using (EFA) is to combine these sub factors according to a theoretical conceptual. Several sub factors and items were dropped and the model's factors were regrouped as TT government support initiatives, transferor characteristics, transferee characteristics, TT environment, and learning centers. In addition, the TT outcome (achievements) factors, are identified and refined, some items were discarded. The outcome of this analysis is the verified model for ITT in ICT projects, which includes a number of refined enabling and achievements variables.

Keywords: International Technology Transfer ITT, Information and Communication Technology ICT, Exploratory Factor Analysis EFA, modelling

1. Introduction

Technology has become one of the essential elements of social and economic development in several newly industrialized countries. Technology is involving knowledge, equipment, and documents can assist firms to upgrade their performance (Wang & Chien 2007). Over the past decade, information and communication technology (ICT) has helped create the most rapidly growing industry sectors, driven efficiency in government and business operations. It has been evidenced by developments from different rising countries that the ICT as a sector can contribute vastly to the national GDP of nations. Furthermore, ICT can efficiently assist international economic integration, narrow the digital divide, and improve living standards. One of the means to impart the advanced technologies to the developing countries is through the Technology transfer process (TT). Several researchers in different industrial areas investigated this process. They were acknowledged that the ITT process is a multifaceted process that is influenced by several factors (Kedia, 1988; Madu, 1989; Cusumano, 1994). The interaction between these factors can affect the level of effectiveness of the TT process. A surveyed literature on relevant TT models showed that none of these researches was dedicated to studying the TT process in the area of ICT-based SME's projects.

Libya, like most developing nations, recognizes the importance of Information and Communication Technology as a catalyst for sustainable socio-economic development. The aim of this study is to develop and empirically tested model that describes the TT process of embracing of foreign advanced technology by ICT companies and ICT-based SME projects in Libya. The variables that extracted and modified from the past investigated studies are classified as factors and sub-factors in a conceptual ICT industry context model. Calantone (1990), Simkoko (1992), Kumar (1999), Lin and Berg (2001), Malik (2002), Wang (2004), Steenhuis (2005), Waroonkun (2008), Mohamed (2010), and khabiri (2012), These relevant models investigation and comparison are presented in

details in previous study of Hassan, A. (2015). The TT model is specially designed to addressing the technology transfer from developed countries to the Libyan ICT industry. Moreover, the results of the evaluation study of the TT process influential and outcome factors in the Libyan ICT industry were gathered from a survey that recently conducted.

1.1 ITT Model Based on Entrepreneur ICT Based SME

The proposed model defines all the important factors that influence the effectiveness of the ITT process and its results (achievements). These relevant factors have been adapted from the examined leading studies into the TT phenomenon with the objective to develop a model that explains the TT process in ICT industry and related SME's projects. Through a process of categorizing variables taken from the previous studies and conceptualizing their relationship with one another in the ICT industry context a number of factors were identified. The identified factors were classified as enabling factors and TT outcome (TT Achievement) factor.

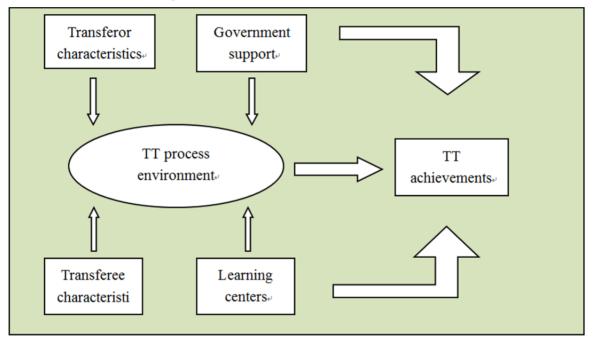


Figure 1. ITT Model for ICT based SME

The developed conceptual model is presented in Figure 1 that illustrates how the developed enabling factors interact to create value (TT achievements) for the host ICT sector. The TT government support initiatives, Transferor characteristics, Transferee characteristics, TT environment, and learning centers are at the left and middle of the model as exogenous constructs, while the outcome factor, TT achievements, has presented as an endogenous construct at the right of the model. The arrows represent the hypothesized causal paths between each enabling and the outcome factor.

The Government support factor was concerned with the degree to which the government's policies and enforcement practices encourage TT to occur. While the learning centers are obviously an important and influential factor in the TT to the developing countries; this factor related to host country-learning capability.

The transferor's characteristic factor is concerned with the transferor's readiness and ability to provide the appropriate technology to the recipient, and transferor's previous experience level in ITT process. On the other hand, the transferee characteristics factor is a significant factor that affects ITT process. Absorptive capacity, prior knowledge, and experience, learning intent, and technological ability identified by the literature as an influential recipient characteristics. The TT process (TT environment) factor explores the effect of technology characteristics, transfer mode, and the relationships between transferor and transferee. The management of TT program and the role of the agent middleman in the TT process are also investigated. The model factors and sub factors listed in Table 1.

code	factor	sub factor
A2.	TT government support initiatives	
A2.1		Government policy (regulations governing the ICT industry).
A2.2		Availability of adequate infrastructure.
A2.3		Government Support.
A2.4		Parent companies encouragement to the skilled workers.
A2.5		ICT Parent companies supporting to ICT SMEs.
B2	Learning centers and ICT entrepreneurs Learning Capability	
B2.1		The educational systems, training programs, and R&D centers.
B2.2		ICT entrepreneurial training and development.
B2.3		ICT Technology based incubator.
B2.4		Involvement of ICT industry in university programs.
C 2	Transferor's characteristic	
C2.1		Transferor's willingness to implement TT initiatives and cooperate with local workers.
C2.2		Transferor's knowledge base and skills.
C2.3		Transferor's ability to transfer technology.
C2.4		Transferor's degree of previous international experience.
D2	Transferee's characteristic	
D2.1		Technology absorption capabilities of the recipient firms.
D2.2		The transferee's degree of experience in ITT process.
D2.3		The shortage of a skilled workforce with the recipient firm.
D2.4		The transferee's motivation to learn new technologies.
E2	TT process. (TT environment)	
E2.1		Complexity level of the technology to be transferred.
E2.2		The mode of technology transfer.
E2.3		The formally planned and well managed TT agreements.
E2.4		The relationship between the transferor and transferee
E2.5		The cultural traits of the both parties.
E2.6		The entrepreneurial agent middleman.

Table 1. Model factors and sub factors

In this study, the performance of and interrelationship between, the above-mentioned TT factors contributes to the achievements of the host ICT sector. The model defines four main areas (sub-factors) where potential benefits derived from international TT initiatives: economic development, project (firm) performance, knowledge and technological capability improvement, and development and survives of ICT technology SME's. The model output factor (TT achievement) explained through four sub-factors and these four sub-factors detailed into several items as shown in Table 2.

code	Factor	Sub factor
A4	Economic development	
A4.1		Host country industrialization and economic development.
A4.2		Local ICT firm's competitiveness in national markets.
A4.3		The financial performance of local ICT firms.
A4.4		Utilization of Libyan natural and human resources.
A4.5		Diversification into new products or markets.
B4	Project (firm) performance	
B4.1		Libyan ICT industry overall long-term performance.
B4.2		Efficiency, services cost and service quality of the host project.
B4.3		Quality standards in Libyan ICT firms.
B4.4		Mastering the new technology, by the Libyan ICT firms.
B4.5		Functional performance of the products, products cost and quality.
C4	knowledge and technological capability improvement	
C4.1		The ICT local firm's technological capabilities and skills base.
C4.2		The recipient's ability to operate, learn new external technologies.
C4.3		Local workers' development.
C4.4		Libyan ICT sector working practices over the long term.
D4	development and survive of ICT technology SME's	
D4.1		Develop and surviving of ICT SMEs.
D4.2		Reducing cost of production, maintain consistency in quality,
		improve productivity for ICT SMEs.
D4.3		The ability to employ a significant amount of the labor.
D4.4		Mastering new process techniques by ICT SMEs.
D4.5		The emergence of ICT entrepreneurs in a small scale enterprise.
D4.6		Increasing technological capabilities and capacities for ICT SMEs.

Table 2. Achievements factors and sub-factors

2. Method

In this research, a quantitative cross-sectional survey was utilized to investigate the TT process in Libyan ICT industry. The questionnaire was sent to the respondents on the second quarter of the year of 2015. Data collection for this study is specifically collected from Libyan ICT industry employees. A questionnaire survey carried out on the sample targeting ICT companies' employees who have involved in TT processes in Libyan ICT industry includes experts, engineers, administrators, technicians, and project supervisors. In total, a 268 questionnaire was distributed and 162 were returned representing a response rate of 60.5 % percent.

The descriptive and inferential statistical techniques are utilized to analyze the quantitative received data to address the research questions. The Statistical Package for the Social Sciences (SPSS 21) software was implemented. The analysis included descriptive statistic to create a respondent profile. Prior to implementation of exploratory factor analysis (EFA), several assumptions were tested: the univariate normality of all factors, anti-image correlation matrix, the factorability, and factors reliability. The Exploratory Factor analysis was performed separately (single factor solution). Such decision to perform a single factor solution was because all the independent variables in this study emerged from a different theoretical background. Therefore, a single factor solution deemed to be appropriate, rather than implementing EFA on all the independent variables simultaneously. The outcome of this analysis is the confirmed model for international TT in ICT projects, which includes a number of refined enabling and achievements variables.

2.1 Research Strategy

The research strategy designed in a way that guarantees accurate execution for each phase of the process. Many forms of research techniques and data collection methods are available to fulfill the desired objectives of this type of study. This research utilizes a quantitative cross-sectional survey to investigate the TT process in Libyan ICT industry; questionnaire surveys can collect a large number of responses within a relatively short time. A questionnaire survey carried out on the sample targeting ICT companies' employees who have involved in TT processes in Libyan ICT industry. Thus, respondents from Libya's ICT industry considered the appropriate

respondents to evaluate the adoptions and the importance of factors affecting the TT process and the outcomes that can gain.

2.2 Sampling Procedure

This study uses the non-probability sampling method instead of using probability sampling. The choice of the non-probability sampling method is due to the absence of the sampling frame of the expert's workforce in Libya who meets the criteria of this study (list of experts who had worked on projects with the association of foreign partners companies are not available). In such case, a convenient sampling method, as part of the non-probability sampling technique, deemed to be an appropriate choice. Since the experts will have the freedom to whether to participate or not in this study.

2.2.1 Sampling Criteria

There are certain criteria of respondent's inclusion that should be followed. Initially, the respondents should possess an adequate number of experiences and have executed a minimum of one TT process projects with foreign companies. Indeed, the sub-factors in the survey are technical. Therefore, it requires the respondents' awareness and vast knowledge and hence the choice of the experts with criteria mentioned above to be strictly followed.

2.2.2 Respondents Selection

The sample of respondents who is able to provide the information needed for the purpose of the study would be selected depends on the experience of the employees of the Libyan ICT sector, particularly technicians, engineers, supervisors, managers involved in TT process in their respective Telecommunication and Information Technology companies.

2.2.3 Sample Size Determination

To find the appropriate sample size for this research, the formula that followed is that formula, which published by the research division of the National Education Association (NEA Research) Bulletin, and accompanied with easy reference table (Krejcie & Morgan 1970).

 $s = X^2 NP (1-P) \div d^2 (N-1) + X^2 P (1-P)$

s = required sample size.

 X^2 = the table value of chi-square for 1 degree of freedom at the desired confidence level

(3.841).

N = the population size.

P = the population proportion (assumed to be .50 since this would provide the maximum x²sample size).

d = the degree of accuracy expressed as a proportion (.05).

3. Results

3.1 Data Screening

The data screening process is conducted to ensure that the data available matches the sampling criteria, guarantee the accuracy of data entry, and specify and treat questionnaires with unanswered items. According to Schafer (1997), responses with more than 5% of missing data should be excluded from further analysis. Besides, responses are excluded based on the sampling criteria guidelines. The received questionnaires were checked against missing data. Few numbers of questionnaires had missing data. These missing data were not more than 5% and was replaced by the mean score (mean substitute method of missing data). As the purpose of this study is to solicit information from an expert who participated in previous TT process with an international partner, it was found that a number of respondents did not indicate any information about participation in TT process; those questionnaires were excluded from further analysis.

Among the 162 received answers, 11 responses were excluded due to that do not adhere with specified sampling criteria such as non-availability of valid experience in TT projects with foreign entities, outliers, missing information, and unengaged response.

3.2 Respondent Profile

Gathering the personal characteristics of the respondents was essential to develop a good understanding of their perspective on the TT process and their field of specialization. The evaluation of the position held by respondents was necessary to confirm the validity and reliability of responses. Determining the experience of

process participants was decisive for ensuring the validity of results. The greater the experience of the respondent in the ICT industry means a higher understanding of process performance and influences.

Respondents were requested to detail their qualifications to confirm they are qualified enough to develop an informed perspective on the ITT process. The aim was to develop a greater understanding of the respondent's exposure to ITT and their experience in the local ICT industry sector. Table 1 summarizes the respondent profile (position- experience -education), according to the position of the respondents results, the majority were engineers with a (41.1%); followed by the technician (17.9%); administrative officer (9.9%); and Project Supervisor (8.6%). The highest frequency of respondents had a bachelor degree (48.3%) while the Diploma holders come with (28.5%) and Master holders percentage (18.5%) followed by the doctorate (4%). Table 3 shows that the respondents have various working experience ranging from less than five years to more than 20 years. The respondents with experience of 6 to 10 years got the highest frequency (39.1%).

Table 3. R	Lespondent Profile
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	Frequency	Percent	Valid Percent	Cumulative Percent
Position				
Manger	8	5.3	5.3	5.3
Project Supervisor	13	8.6	8.6	13.9
Consultant	6	4.0	4.0	17.9
Academic Staff	10	6.6	6.6	24.5
Engineer	62	41.1	41.1	65.6
Technician	27	17.9	17.9	83.4
Administrative Officer	15	9.9	9.9	93.4
Others	10	6.6	6.6	100.0
Total	151	100.0	100.0	
Experience				
less than five years	27	17.9	17.9	17.9
6-10 years	59	39.1	39.1	57.0
11-15 years	35	23.2	23.2	80.1
16-20	20	13.2	13.2	93.4
more than 20 years	10	6.6	6.6	100.0
Total	151	100.0	100.0	
Education				
Diploma	43	28.5	28.5	28.5
Bachelor	73	48.3	48.3	76.8
Master	28	18.5	18.5	95.4
Doctorate	6	4.0	4.0	99.3
Other	1	.7	.7	100.0
Total	151	100.0	100.0	

3.2 Exploratory Factor Analysis (EFA)

The purpose of implementing exploratory factor analysis (EFA) is to reduce or trim the number of sub-factors, to establish reliable and validated measures for the factors under investigation of this study. Furthermore, the goal of using EFA is to combine the sub-factor according to a theoretical conceptual. However prior to implementation of exploratory factor analysis (EFA), several assumptions were checked. As suggested by Hair et al. (2006) several assumptions to be tested, before the implementation of EFA. Such assumptions are the univariate normality of all sub-factors, anti-image correlation matrix. The sub-factors should have factorability (adequacy of the sample size to perform EFA), which is measured by KMO (Kaiser – Mayer – Olkin) and Barlett test.

Indices	Measured by	Recommended range	references
Sample size		> 100	Hair et al. (2010); and Mundfrom, (2005)
Univariate normality	Skewness and Kurtosis	absolute 2	Hair et al. (1998)
Factorability.	anti-image correlation	All the diagonal elements > 0.5	Hair et al. (2006).
	matrix	and the off- diagonal elements have low correlation.	Williams, Brown & Onsman (2012)
	KMO (Kaiser – Meyer Olkin)	≥ 0.5	Kaiser (1974)
Emerging factors	Eigenvalues	> 1	Garson, 2010
Factor loadings		> 0.5	Charles Zaiontz. (2013)
Reliability	Cronbach's Alpha	> 0.60	Hair et al., 2010

Table 4. Indices index

3.2.1 TT Government Support Initiatives

Reflecting the assumptions mentioned above, it was found that all sub-factors related to the TT government support initiatives satisfy the condition of univariate normality, as all the sub-factors has skewness and kurtosis fall within the recommended range to indicate normality. As advised by Hair et al. (1998), the recommended value is absolute two for both skewness and kurtosis. Table 5 displays the corresponding skewness and kurtosis values.

Table 5. Descriptive Statistics

	Ν	Skewness		Kurtosis	
	Statistic	Statistic	Std. Error	Statistic	Std. Error
A2.1	151	-1.158	.197	.857	.392
A2.2	151	-1.004	.197	.643	.392
A2.3	151	574	.197	359	.392
A2.4	151	961	.197	.776	.392
A2.5	151	653	.197	022	.392

Furthermore, to ensure the adequacy of sample size, the associated KMO test was examined. Kaiser (1974) recommended the value of KMO to be above 0.5 and Bartlett's Test of Sphericity to be significant (P-value <0.05), to make the statistical inference that there is factorability. With inspection of Table 6, the KMO test reported a value of 0.767, which is above the recommended threshold of 0.5, additionally, the Bartlett's Test of Sphericity indicated a significant P value (Sig. = 0.000). Such statistical values satisfy the condition of EFA implementation, and the data, therefore, possess factorability.

Table 6. KMO and Bartlett's Test

Kaiser-Meyer-Olkin	Measure of Sampling Adequacy.	.767
	Approx. Chi-Square	237.363
Bartlett's Test of Sphericity	df	10
	Sig.	.000

Furthermore, the anti-image correlation table depicted in Table 7 showed the satisfaction of EFA implementation conditions; that is all the diagonal elements should exceed the value of 0.5, and the off- diagonal elements have a very low correlation, as recommended by Hair et al. (2006). Therefore, it could be said that implementing EFA on the items related to TT Government Support Initiatives is justified.

m 1 1			
Table	/ An	ti-imag	ge Matrix
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	A2.1	A2.2	A2.3	A2.4	A2.5
A2.1	.863 ^a	153	216	075	084
A2.2	153	.818 ^a	036	353	024
A2.3	216	036	.839 ^a	068	270
A2.4	075	353	068	.709 ^a	569
A2.5	084	024	270	569	.725 ^a

a. Measures of Sampling Adequacy(MSA)

a

Since the assumptions of EFA were met, the implementing of exploratory factor analysis (EFA) with the use of varimax principal component method had resulted into a single factor solution, as it was theorized. Since this factor represents a single dimension, therefore, the theoretical expectation would suggest an emerging of single factor solution.

The inference of single factor solution was detected through inspecting the eigenvalues. The eigenvalues that are above 1 is an indication of emerging factors. Only one factor has emerged since the eigenvalues presented in Table 8 has only one eigenvalue above 1 that is equal to 2.764. The emerged single factor was able to explain 55.279 percent of the variance. This means that all the five sub-factors were able to explain 55.279 percent of the variation to the corresponding factor. Totally explained the variance of little as 50 percent is deemed acceptable, especially for a single factor solution (Garson, 2010).

Table 8. Total Variance Explained

Compo	onentInitial E	Eigenvalues		Extraction	on Sums of Squared	Loadings
	Total	% of Variance	Cumulative %	Total	% of Variance	Cumulative %
1	2.764	55.279	55.279	2.764	55.279	55.279
2	.760	15.196	70.475			
3	.690	13.808	84.283			
4	.521	10.421	94.704			
5	.265	5.296	100.000			

Extraction Method: Principal Component Analysis.

On the other hand, all the sub-factors has adequate loading that fall within the recommended range of above 0.5, as suggested by Hair et al. (2006). Therefore, the analysis will further proceed with all sub-factors related to the TT government support initiatives, without any further exclusion. The factor loadings are reflected in Table 9 and they are ranging from 0.628 until 0.848, and the reliability test shows that the factor has a reliability of 0.794.

	Component	
A2.1	.628	
A2.2	.695	
A2.3	.689	
A2.4	.848	
A2.5	.833	
Reliability Statistic	s	
Cronbach's Alpha	.794	
N of Items	5	

Table 9. Component Matrix ^a

Extraction Method: Principal Component Analysis.

a. 1 components extracted.

3.2.2 Learning Centers and ICT entrepreneurs Learning Capability

The same assumptions before the EFA implementation were checked against the sub-factors related to the learning ICT entrepreneurs Learning Capability. As previously mentioned, EFA assumptions were investigated in terms of univariate normality. As shown in Table 10, the skewness and kurtosis values for all sub-factors related the learning centers and ICT entrepreneurs learning capability fall within the recommended range to indicate normality; all the values did not exceed absolute 2 of skewness and kurtosis.

Table 10. Descriptive Statistics

	Ν	Skewness		Kurtosis	
	Statistic	Statistic	Std. Error	Statistic	Std. Error
B2.1	151	-1.352	.197	1.285	.392
B2.2	151	906	.197	.597	.392
B2.3	151	766	.197	.288	.392
B2.4	151	985	.197	.281	.392

Furthermore, KMO test was performed to judge on the sample adequacy of the learning centers and ICT entrepreneurs learning capability data. The results displayed in Table 11 indicate that the sample size of these data was adequate. In addition, Bartlett's Test of Sphericity was significant, which indicates that the use of EFA is appropriate.

Table 11. KMO and Bartlett's Test

Kaiser-Meyer-Olkin Measure of Sampling Adequacy775			
	Approx. Chi-Square	318.837	
Bartlett's Test of Sphericity	df	6	
	Sig.	.000	

The last assumptions of the successful implementation of EFA were investigated through the anti-image correlation matrix. As displayed in Table 12, all the diagonal elements of the anti-image matrix have the value of above 0.5 and off-diagonal element with weak correlations.

Table 12. Anti-image Matrix ^a

B2.1 B2.2 B2.3 B2.4
B2.1.738 ^a 500447.087
B2.2500.797 ^a 210182
B2.3447210.774 ^a 392
B2.4.087182392.804 ^a

a. Measures of Sampling Adequacy(MSA).

By using the principle component analysis (CPA) with a use of the varimax rotational method, a single factor solution was extracted, as indicated by the eigenvalue score shown in Table 13. Such a single factor solution matches the theoretical conceptual that the sub-factors represent one dimension. Moreover, the four sub-factors used to represent the learning centers and ICT entrepreneurs learning capability was explaining 71.655 per cent of variation, which deemed to be excellent and exceeds the minimum cut-off 50 percent. The use of single factor solution is justified, since the learning centers and ICT entrepreneurs learning capability theoretically a distinctive dimension that has no theoretical relationship with other factors in this study, hence, the use of the single-factor solution is deemed an appropriate choice.

ComponentInitial Eigenvalues			Extraction St	ums of Squared Lo	oadings	
	Total	% of Variance	Cumulative %	Total	% of Variance	Cumulative %
1	2.866	71.655	71.655	2.866	71.655	71.655
2	.615	15.378	87.033			
3	.300	7.493	94.526			
4	.219	5.474	100.000			

Table 13. Total Variance Explained

Extraction Method: Principal Component Analysis.

In addition, with reference Table14, all the sub-factors have adequate factor loadings that exceed the threshold of 0.5, and the reliability test shows that the factor has a reliability of 0.862. Accordingly, the analysis will proceed with all sub-factors as they do not pose any threat to the validity at the later stage of analysis.

Table 14. Component Matrix ^a

	Component
	1
B2.1	.874
B2.2	.877
B2.3	.897
B2.4	.727
Reliability Statistics	
Cronbach's Alpha	.862
N of Items	4

Extraction Method: Principal Component Analysis.

a. 1 components extracted.

3.2.3 Transferor's Characteristics

The same assumptions were investigated. Table 15 reflects that the sub-factors of Transferor's characteristics have skewness and kurtosis within the recommended range, which further indicate that all sub-factors belong to this factor possess univariate normality.

Table 15. Descriptive Statistics

	Ν	Skewnes	S	Kurtosis	
	Statistic	Statistic	Std. Error	Statistic	Std. Error
C2.1	151	714	.197	.289	.392
C2.2	151	452	.197	898	.392
C2.3	151	758	.197	.383	.392
C2.4	151	816	.197	.582	.392

Both conditions of sample adequacy and anti-image correlation were satisfied and, therefore, the implementation of a single factor solution is permissible. Such results are displayed respectively in Table 16 and 17.

Table 16. KMO and Bartlett's Test

Kaiser-Meyer-Olkin Measure of Sampling Adequacy731			
	Approx. Chi-Square	91.656	
Bartlett's Test of Sphericity	df	6	
	Sig.	.000	

Table 17. Anti-image Matrix ^a

C2.1 C2.2 C2.3 C2.4
C2.1.749 ^a 234217136
C2.2234.735 ^a 121262
C2.3217121.729 ^a 289
C2.4136262289.715 ^a

a. Measures of Sampling Adequacy(MSA).

Additionally, as the theory was stated, a single factor emerged from the EFA implementation. The sub-factors related to Transferor's characteristics were able to explain 51.950 percent of the factor, as shown in Table 18. Such total variance is sufficient and meets the minimum threshold of acceptable total variance explained.

Table 18. Total Variance Explained

Component	Initial Eigenvalues			Extraction Sums of Squared Loadings		
	Total	% of Variance	Cumulative %	Total	% of Variance	Cumulative %
1	2.078	51.950	51.950	2.078	51.950	51.950
2	.697	17.436	69.386			
3	.673	16.828	86.215			
4	.551	13.785	100.000			

Extraction Method: Principal Component Analysis.

All the corresponding sub-factors of organizational performance showed an appropriate loading that fall within the range of recommended (above 0.5). Hence, no sub-factors were deleted for subsequent analysis. Such factor loadings are reflected in Table 19 and ranging from .700 until .746, and the reliability test shows that the factor has a reliability of 0.691.

Table 19. Component Matrix ^a

	Component
	1
C2.1	.700
C2.2	.716
C2.3	.721
C2.4	.746
Reliability Statistic	s
Cronbach's Alpha	.691
No of Items	4

Extraction Method: Principal Component Analysis.

a. 1 components extracted.

3.2.4 Transferee's Characteristic

As indicated in Table 20, all the sub-factors has skewness and kurtosis within the recommended range of not more than absolute 2.

Table 20. Descriptive Statistics

Ν	Skewn	Skewness		is
Statis	ticStatist	icStd. Er	rorStatisti	cStd. Error
D2.1151	775	.197	.330	.392
D2.2151	224	.197	878	.392
D2.3151	505	.197	217	.392
D2.4151	757	.197	.019	.392

The initial inspection of the EFA assumptions related to the sub-factors of Transferee's characteristic impact yielded undesirable statistical inference. As KMO test was on the borderline, as well as the total variance explained was below the minimum acceptable value of 50 per cent. Accordingly, the sub-factors with lowest factor loading (D2.3) were removed from the analysis, in the attempt to improve the overall result and satisfy the EFA conditions. Subsequent to the removal of the D2.3, the results show improvement. This demonstrated by the KMO test that had improved to 0.597, with significant Test of Sphericity. Furthermore, the diagonal element of anti-image correlation has the value of above 0.5, as shown in Table 21 & 22, respectively.

Table 21. KMO and Bartlett's Test

Kaiser-Meyer-Olkin Measure of Sampling Adequacy597				
	Approx. Chi-Square	55.527		
Bartlett's Test of Sphericity	df	3		
	Sig.	.000		

Table 22. Anti-image Matrix ^a

D2.1 D2.2 D2.4
D2.1.597 ^a 076409
D2.2076.658 ^a 297
D2.4409297.568 ^a

a. Measures of Sampling Adequacy(MSA).

The total variance explained, which was the main cause of the problem has improved after removing D2.3. Initially, the total explained variance was 43.42 percent; however, it was improved to 56.711 percent, which exceeds the threshold. This statistical figure is present in Table 23. Even with removing D2.3, a single factor emerged, whereby the retained sub-factors were able to explain 56.711 percent. The eigenvalues score indicate the single factor solution, which there was only one value that exceeds 1 and it was equal to 1.701.

Table 23. Total Variance Explained

Component	Initial Eig	envalues		Extraction Sums of Squared Loadings				
	Total	% of Variance	Cumulative %	Total	% of Variance	Cumulative %		
1	1.701	56.711	56.711	1.701	56.711	56.711		
2	.782	26.063	82.773					
3	.517	17.227	100.000					

Extraction Method: Principal Component Analysis.

The three retained sub-factors have factor loadings that exceed the threshold of 0.5. As shown in Table 23, the factor loadings are ranging from 0.752 until 0.830, and the reliability test shows that the factor has a reliability of 0.611.

Table 23. Component Matrix ^a

	Component
	1
D2.1	.752
D2.2	.669
D2.4	.830
Reliability Statistics	
Cronbach's Alpha	.611
N of Items	3

3.2.5 TT process. (TT environment)

Table 25 reflects the skewness and kurtosis for the sub-factors related to TT process factor. The condition of univariate normality is satisfied through the inspection of skewness and kurtosis values, which fall within the recommended range.

Table 25. Descriptive Statistics

	Ν	Skew	ness	Kurtosis	
	Statistic	Statistic	Std. Error	Statistic	Std. Error
E2.1	151	536	.197	184	.392
E2.2	151	310	.197	534	.392
E2.3	151	371	.197	294	.392
E2.4	151	869	.197	.433	.392
E2.5	151	554	.197	007	.392
E2.6	151	415	.197	445	.392

The initial inspection of the single factor solution has resulted in poor total variance explained. Therefore, there was a need to improve the statistical figures. Such improvement is achieved by removing sub-factors with low loading. The inspection of factor loadings, suggested the removal of E2.4 & E2.6, as they had the lowest factor loadings.

With such removal, the assumption of EFA was revisited. The KMO and Bartlett's Tests in Table 26 shows that the condition of factorability is attained, as KMO test has the value of 0.737 that is above the threshold of 0.5 and the Bartlett's Test is significant.

Table 26. KMO and Bartlett's Test

Kaiser-Meyer-Olkin Measure of Sampling Adequacy737					
	Approx. Chi-Square	103.208			
Bartlett's Test of Sphericity	df	6			
	Sig.	.000			

In addition, the anti-image matrix has diagonal elements of the values of more than 0.5, as seen in Table 27.

Table 27. Anti-image Matrix ^a

a. Measures of Sampling Adequacy(MSA).

With the deletion of E2.4 and E2.6, the total variance explained has been further improved. It was improved to 52.238 per cent, before the improvement made, the total explained variance was merely 40 percent. Yet, the retained four sub-factors were able to extract a single factor solution with total variance explanation of 52.238 per cent. Such statistical figures are presented in Table 28.

Component	Initial Ei	genvalues		Extraction Sums of Squared Loadings			
	Total	% of Variance	Cumulative %	Total	% of Variance	Cumulative %	
1	2.130	53.238	53.238	2.130	53.238	53.238	
2	.744	18.608	71.846				
3	.619	15.487	87.333				
4	.507	12.667	100.000				

Table 28. Total Variance Explained

Extraction Method: Principal Component Analysis.

The retained four factors were further checked in terms of their factor loadings. As shown in Table 29, all the retained sub-factors have the recommended value of factor loadings that exceeds the cutoff 0.5, and the reliability test shows that the factor has a reliability of 0.698.

Table 29. Component Matrix ^a

	Component 1
E2.1	.765
E2.2	.796
E2.3	.663
E2.5	.686
Reliability Statistics	
Cronbach's Alpha	.698
No of Items	4

Extraction Method: Principal Component Analysis.

a. 1 components extracted.

3.2.6 Achievements

First, the sub-factors from A4.1 until D4.6 showed adequate univariate normality. The univariate normality was judged based on the corresponding values of skewness and kurtosis. As Table 30 indicated that all the skewness and kurtosis fall within the recommended range advised by Hair et al. (1998) to indicate that the items have univariate normality, the recommended value is absolute 2 for both skewness and kurtosis.

Table 30. Descriptive Statistics

	Ν	Skewness		Kurtosis	
	Statistic	Statistic	Std. Error	Statistic	Std. Error
A4.1	151	368	.197	938	.392
A4.2	151	455	.197	541	.392
A4.3	151	316	.197	503	.392
A4.4	151	568	.197	171	.392
A4.5	151	839	.197	.589	.392
B4.1	151	376	.197	791	.392
B4.2	151	310	.197	707	.392
B4.3	151	586	.197	438	.392
B4.4	151	761	.197	.124	.392
B4.5	151	628	.197	286	.392
C4.1	151	461	.197	746	.392
C4.2	151	363	.197	800	.392
C4.3	151	504	.197	343	.392
C4.4	151	519	.197	672	.392
D4.1	151	554	.197	288	.392
D4.2	151	523	.197	404	.392
D4.3	151	723	.197	.495	.392
D4.4	151	728	.197	.248	.392
D4.5	151	841	.197	.835	.392
D4.6	151	568	.197	.143	.392

Secondly, as per requirement of the performing EFA, Hair et al. (1998) suggested that KMO test should have the value of above 0.5, which then indicate that the sample is adequate to perform EFA. As shown in Table 31, the KMO test has the value of 0.931, which indicates that the sample size is adequate to perform EFA.

Table 31. KMO and Bartlett's Test

Kaiser-Meyer-Olkin Measure of Sampling Adequacy931						
	Approx. Chi-Square	1473.963				
Bartlett's Test of Sphericity	df	120				
	Sig.	.000				

Furthermore, the anti-image correlation table depicted in Table 32 showed the satisfaction of EFA implementation conditions; that is all the diagonal elements should exceed the value of 0.5, and the off-diagonal elements have a very low correlation, as recommended by Hair et al. (2006). Therefore, it could be said that implementing EFA on the sub-factors related to achievements is justified. On the hand, the total variance explained was again lower than the minimum acceptable percent; this necessitates removing some sub-factors to improve the total variance explained. Several sub-factors with lowest factor loadings were detected, and they were removed, such sub-factors are B4.2, C4.4, D4.1, and D4.2. This left the achievement to present with 16 sub-factors, instead of 20 sub-factors.

The retained 16 sub-factors were used for further analysis. As shown in Table 33, the 16 retained sub-factors were able to generate a single factor solution for the achievement with an appropriate total explained variance of 53.195 per cent. The single factor solution was indicated by the eigenvalue score of above 1. Moreover, all the retained 16 sub-factors have factor loadings that exceed the recommended value of 0.5, and the reliability test shows that the factor has a reliability of 0.698 as shown in Table 34.

	A4.1	A4.2	A4.3	A4.4	A4.5	B4.1	B4.3	B4.4	B4.5	C4.1	C4.2	C4.3	D4.3	D4.4	D4.5	D4.6
A4.1	.944ª	239	130	055	082	259	.154	089	087	.041	055	.032	.054	066	.005	067
A4.2	239	.951ª	158	054	137	037	082	032	.074	261	013	021	154	.106	041	106
A4.3	130	158	.909ª	212	013	.036	068	.134	148	.056	035	100	010	.073	.133	171
A4.4	055	054	212	.912ª	333	.001	.015	.011	.025	.024	110	078	200	.037	189	.198
A4.5	082	137	013	333	.931ª	.021	045	252	005	.020	.080	034	.200	117	.000	199
B4.1	259	037	.036	.001	.021	.935ª	371	.002	009	199	.045	.037	.026	.039	092	040
B4.3	.154	082	068	.015	045	371	.926ª	170	280	056	222	.081	102	.077	.004	055
B4.4	089	032	.134	.011	252	.002	170	.932ª	321	089	.055	305	018	.036	178	.008
B4.5	087	.074	148	.025	005	009	280	321	.937ª	103	055	.077	.118	120	091	.050
C4.1	.041	261	.056	.024	.020	199	056	089	103	.937ª	137	225	007	231	.189	060
C4.2	055	013	035	110	.080	.045	222	.055	055	137	.950ª	207	.068	166	122	.041
C4.3	.032	021	100	078	034	.037	.081	305	.077	225	207	.931ª	102	032	.010	.096
D4.3	.054	154	010	200	.200	.026	102	018	.118	007	.068	102	.909ª	346	145	119
D4.4	066	.106	.073	.037	117	.039	.077	.036	120	231	166	032	346	.925ª	088	239
D4.5	.005	041	.133	189	.000	092	.004	178	091	.189	122	.010	145	088	.931ª	351
D4.6	067	106	171	.198	199	040	055	.008	.050	060	.041	.096	119	239	351	.924ª

Table 32. Anti-image Correlation Matrices

a. Measures of Sampling Adequacy(MSA)

Component	Initial	Eigenvalues		Extraction Sums of Squared Loadings				
	Total	% of Variance	Cumulative %	Total	% of Variance	Cumulative %		
1	8.511	53.195	53.195	8.511	53.195	53.195		
2	.972	6.073	59.268					
3	.935	5.842	65.110					
4	.825	5.154	70.264					
5	.696	4.350	74.615					
6	.602	3.761	78.376					
7	.544	3.401	81.776					
8	.501	3.130	84.906					
9	.434	2.712	87.618					
10	.423	2.643	90.261					
11	.367	2.294	92.555					
12	.309	1.929	94.484					
13	.243	1.522	96.005					
14	.227	1.419	97.424					
15	.214	1.338	98.762					
16	.198	1.238	100.000					

Table 33. Total Variance Explained

Extraction Method: Principal Component Analysis.

Table 34. Component Matrixa

	Component 1
A4.1	.701
A4.2	.795
A4.3	.532
A4.4	.658
A4.5	.753
B4.1	.726
B4.3	.791
B4.4	.826
B4.5	.755
C4.1	.785
C4.2	.711
C4.3	.674
D4.3	.650
D4.4	.749
D4.5	.755
D4.6	.756
Reliability Statistics	
Cronbach's Alpha	.941
No of Items	16

Extraction Method: Principal Component Analysis.

a. 1 components extracted.

4. Discussion

The ICT international technology transfer model was developed, and the affecting factors and sub factors were explored and defined. The developed model was empirically tested in Libyan ICT industry; a questionnaire survey was carried out on the sample targeting ICT companies' employees who have involved in TT processes. Respondents were requested to provide a rating for their opinion to the factors and their perceived impact on a TT process. The Statistical Package for the Social Sciences (SPSS 21) software was implemented to analyze the receiving data. The analysis included descriptive statistic to create a respondent profile. To minimize and establish reliable and validated measures for the factors under investigation of this study Exploratory Factor

Analysis (EFA) was implemented. Furthermore, the goal behind using (EFA) is to combine these sub-factors according to a theoretical conceptual. Several assumptions were tested prior to implementing the exploratory factor analysis (EFA): the univariate normality of all factors, anti-image correlation matrix, the factorability, and factors reliability. All the independent variables in this study emerged from a different theoretical background. Therefore, a single factor solution deemed to be appropriate, rather than implementing EFA on all the independent variables simultaneously.

The outcome of this analysis is the confirmed model for international TT in ICT projects, which includes a number of refined enabling and achievements variables. Among five enablers (factors) and their 23 sub-factors, the analysis retains only 20 and ignores 3 sub factors. The (EFA) grouped the following factors (enablers) in the developed model: TT government support initiatives, transferor characteristics, transferee characteristics, TT environment, learning centers, and their respective sub-factors (variables). On the hand, the outcome (achievements) construct has resulted with only 16 sub factors and 4 sub factors were ignored. This paper is a part of an ongoing study to develop a confirmed technology transfer model for the Libyan ICT industry. Utilizing the confirmatory factor analysis (CFA), and structural equation modeling (SEM) techniques will be the next phase to achieve more confirmation and validation of the model's remained factors and sub factors and to confirm the model structure and causal paths between factors.

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