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# **Development of Structural Mediation Model of Innovative** Strategy, Employees' Job Performance and Technology for **UAE Government Sector**

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**Abstract:** This study established a relationship of innovations and employees' job performance in the government sector and also the mediating effect of technology towards the relationship. It adopted quantitative approach where data was collected through questionnaires survey and analyzed statistically. The questionnaire survey utilized simple random sampling methods among UAE government employees. A total of 300 questionnaires were distributed randomly and 265 were returned indicates 88.3% response rate. The collected data was then used to develop and assess SEM-AMOS mediation model. The results found that all the direct relationships reported a positive effect. It also found that the innovation dimension constructs and technology mediating construct having explained 84% variation in employees' job performance constructs. For indirect/mediation effect, it was found that technology does not show any mediation effect on the relationship between innovation and job performance constructs. This implies that technology does not mediate the relationship between innovation strategies and the employees' job performance. It also implies that for an effective employees' job performance core attention to building and managing an innovation approach is required in the UAE government sector. Thus, innovation strategies give more influence to employees' job performance in the government sector through the distinctive capabilities of UAE ministry employees.

**Keywords:** Innovation, job performance, technology

# 1. Introduction

Driven by global competition to improve service delivery, the UAE government in latest decade has initiated the transformation of the public sector to assist the country as a center for business excellence in the Middle East. The public sector needs continuous improvement in the employees' performance to deliver high-quality services as stated by Suliman & Alkathairi (2013). Besides that the sector has to overcome the challenges faced on the existing practices and reviewing the performance measurement system of the public sector administration (Almansoori, 2011; Turkyilmaz et al., 2011). It also needs to address problems like lack of effective and efficient employees' performance;

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lack of creativity components; innovation, competitiveness and profitability in the UAE public sector (Suliman & Alkathairi, 2013; Mohammed et al., 2018). There many studies on employee's job performance in the public sector but are concentrated on management commitment and hygiene factors (Suliman & Akathairi, 2013); performance assessment (Almansoori, 2011); employee engagement and loyalty (Ibrahim & Falasi, 2014); employees' remuneration and incentives (Al Naqbi et al., 2018).

However, there is a lack of study on the effect of innovation to employee job performance where this study was intended to fill. Hence this study was to establish the relationship between innovation strategies with employees' job performance for the UAE government sector. The innovation components used in this study are adopted from Oslo Manual developed by the OECD and the European Commission in 2005, which describes four types of innovations namely product and services innovations, process innovations, organizational innovation, and marketing innovations. While for job performance factors, it could be seen from many perspectives but this study considered the dimension of job performance into four groups namely task performance, contextual performance [interpersonal], contextual performance [organizational] and adaptive performance (Gunday et al., 2011; Weisburd & Braga, 2019). The study further explored technology as the mediating role on the relationship between innovation approach and employee job performance (Rosrofli & Kamal Ariffin, (2014); Hamidu et al., (2015); Fagot-Bouquet et al., 2016) which made the study novel.

## 2. Methodology

Literatures provide evidences that innovation drives from the human resource practices in which employee behavior and attitude influenced the innovation (Awan et al, 2015). Variety of human resource practices support the employee to get innovative or competences. However, employees behave corresponding with the way the organization shows commitment to them as individual. Therefore, when organization wants employees to work innovatively then it needs to focus on creating commitment to the workforce (Awan et al, 2015). This is because innovation is an important way to overcome the challenges in the current rapidly changing environment and employees highly influence the innovation performance of organization.

This study was applied to the UAE government sector to ascertain the Role of Innovation Strategies on Employees' Job Performance. A total of 300 structured questionnaires were distributed to the sample of the Ministry of Culture, Youth and Community Development of UAE staffs to find out the effects of innovations on employees; job performance of which 260 responses were found valid. The collected data were used in the development of Analysis of Moment Structures Equation Modelling (AMOS-SEM) model which was based on the framework model and the hypotheses postulated for this research as Fig 1.

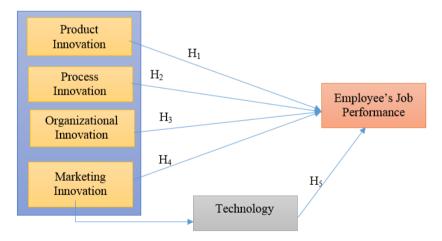


Fig 1 - Conceptual model

The framework model of figure 1 consists of innovation strategy having four independent constructs which are product innovation, process innovation, organizational innovation and marketing innovation; one mediating construct which is technology and one dependent construct which is employee's job performance. The hypotheses of this model are as follow;

- H1: Product innovation has a positive relationship with job performance
- H2: Process innovation has a positive relationship with job performance
- H3: Organizational innovation has a positive relationship with job performance
- H4: Marketing Innovation has a positive relationship with job performance
- H5: Technology Mediate the relationship of innovation strategy and job performance

#### 3. Measurement Model Assessment

The assessment of measurement model is conducted through Confirmatory factor analysis (CFA) using AMOS-SEM software. It was conducted by adopting the recommended CB-SEM methodological process found in the multivariate texts (Hair, et al, 2010; Byrne, 2010; Kline, 2011). The analysis was conducted by specifying the model, model identification, parameter estimation, assessment of model's goodness-of-fit and lastly, re-specification of the model. Additionally, at each stage of the analysis, the validity of the models was evaluated based on the established criteria for CB-SEM evaluation presented in Table 1.

Table 1 - Recommended Goodness-of-fit Indices and acceptance levels
(Awang, 2012; Kline, 2011; Byrne, 2010)

<b>Index Category</b>	Indices	Acceptable level
Absolute fit	Chisq.	P < 0.05
Absolute fit	<b>RMSEA</b>	Value < 0.08
Absolute fit	GFI	Value > 0.90
Incremental fit	AGFI	Value > 0.90
Incremental fit	CFI	Value > 0.90
Parsimonious fit	Chisq./df	Value < 5.0

Then maximum likelihood estimation (MLE) method was used to estimate the parameters of a statistical model given observations by finding the parameter values that maximize the likelihood of making the observations given the parameters. Hence, the constructs in this framework model which are the measurement and structural models were evaluated according to the established goodness-of-fit indices as presented in Table 1.

#### 3.1 Assessment of individual construct

In this study the measurement model comprises of 9 individual constructs namely product innovation; process innovation; organizational innovation; marketing innovation; task performance; contextual performance interpersonal; contextual performance organizational; adaptive performance and technology. For product innovation (PI), the measurement model was constructed using seven (7) indicators which represented the measurement items for the construct. Using the AMOS graphic, the validity of the measurement model was assessed by examining the factor loadings, the squared multiple regression (R2) and the fitness indices. For a model to be considered as valid, the literature suggests that factor loading and R2 should not be less than 0.50 and 0.30 respectively (Kline, 2011; Byrne, 2010). Hair et al (2011) suggested that any factor/indicator having values less than the recommended minimum should be deleted from the measurement model. In line with the prescribed procedure, the measurement model for Product Innovation was evaluated as figure 1 and found that the seven indicators values within the recommended thresholds. However, examination of the goodness-of-fit indices revealed that the measurement model does not meet the requirement for model acceptance. Specifically, the RMSEA and Chi sq/df reported values far greater than the recommended thresholds. Similarly, both NFI and GFI reported values less than the recommended .90, hence indicating that model re-specification is required.

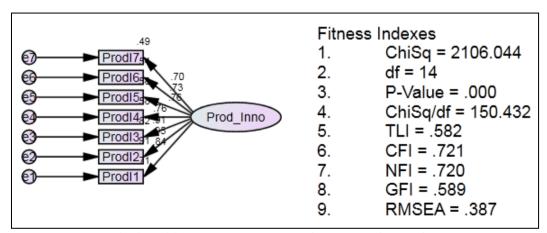


Fig. 2 - Initial measurement model for product innovation

Fig. 2 shows the re-specified measurement model for product Innovation. Upon freeing up some parameters in the initial model by co-varying two error terms e3 and 36, the desired model fitness is achieved. All the model fitness indices returned values within the acceptable thresholds. RMSEA reported .078, Chi sq/df=2.612 while GFI and NFI

reported values greater than the recommended value of .90. Thus, the final model is accepted and retained for inclusion in the structural model.

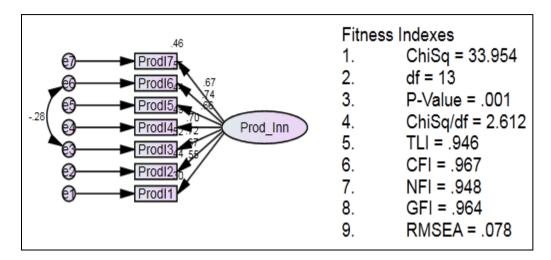


Fig. 3 - Final measurement model Product Innovation

The same procedures were conducted for the other eight individual constructs. Having established the validity and model of fit of the individual constructs, it is necessary to assess the validity of the entire measurement model of the research prior to evaluating the structural model. The rationale for conducting this analysis was to establish the validity of all the constructs in unison. The validity of the entire measurement model was assessed by examining the convergent validity and discriminant validity of the constructs in the model.

#### 3.2 Assessment of Entire Measurement Model

According to Hair et al (2010), two parameters can be used which are convergent validity and discriminant validity in assessing the entire measurement model. Convergent validity is a measure of the degree to which the items or indicators of a construct are correlated with the construct. While discriminant validity measures the degree to which a construct is distinct from other constructs in the model.

In CFA SEM analysis, the convergent validity of construct is assessed by Bentler\_Bonett coefficient (NFI). Recommended threshold for convergent validity using the NFI index is 0.90 (Hair, et al, 2010; Kline, 2011; Byrne, 2010). Using the factor loading and the NFI criteria, the convergent validity of the individual final measurement models is as in Table 2.

Table 2 - Convergent validity measures of the entire final models

Nie	Construct	Residual items	Factor	NIET I. do.	
No		number	Lowest FL	Highest FL	NFI Index
1	Product Innovation	7	.546	.741	.948
2	Process Innovation	4	.558	.778	.981
3	Organisational Innovation	5	.634	.742	.977
4	Marketing Innovation	7	.676	.798	.961
5	Task Performance	5	.871	.988	.999
6	Contextual Performance Interpersonal	4	.784	.955	.998
7	Contextual Performance organizational	4	.699	.917	.990
8	Adaptive Performance	6	.676	.793	.967
9	Technology	7	.696	.813	.967

Table 2 indicates that all constructs satisfy the acceptable threshold. For discriminant validity, it is achieved when the squared inter-construct correlations associated with a particular construct are greater than the corresponding inter-construct correlation estimates with other constructs (Hair et al, 2010). The recommended threshold for AVE is  $\geq$ 0.50 (Hair, et al. 2014) and the discriminant validity of this study is as in table 3.

	JP	Pro_Inn	Mkt_Inn	Proc_Inn	Org_Inn	Techno
JP	0.781					
Pro_Inn	0.006	0.843				
Mkt_Inn	-0.010	0.815	0.839			
Proc_Inn	-0.239	0.708	0.704	0.792		
Org_Inn	0.340	0.619	0.662	0.726	0.803	
Techno	0.581	-0.253	-0.126	0.065	0.409	0.844

Table 3 shows the AVE of each construct at the diagonal while the off-diagonal values represent the correlation coefficients between the constructs. Based on the recommended threshold, all the AVEs are greater than 0.50 and each AVE value is higher than any correlation with other construct, hence indicating the achievement of discriminant validity.

## 4. Structural Model Evaluation

Once the entire measurement model has achieved the assessment criteria, the model can then be assessed at the structural component. The following stage is to conduct path analysis to evaluate the structural part of the model. It assessed the mediating relationship that is the exogenous variables which are product innovation, market innovation, process innovation and organizational innovation with the mediator variable which is the technology and the exogenous variable which is the job performance.

At the initial assessment of the structural model, it was found that some indexes failed to meet the acceptable level. All the observed factor loadings and their corresponding square multiple regression meet the required thresholds of 0.50 and 0.30 respectively. In respect of the fit indexes, the RMSEA, CFI, and GFI do not satisfy the criteria for acceptance while only the p-value reported values within the acceptable limit. This suggests that model re-specification was required. Then re-specification of the model was carried out by freeing off some parameters through co-variation and the results are as in Fig 4

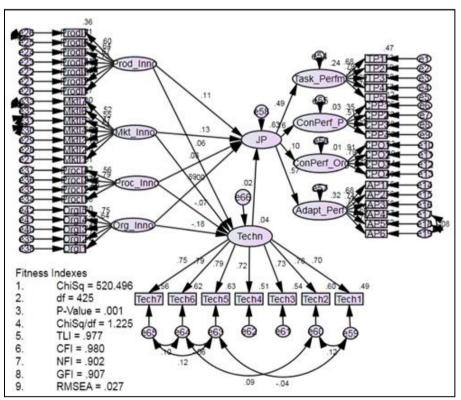


Fig. 4 - Final structural model

At structural modelling in CFA, the model has to meet up with the criteria for model fit. However, Hair et al (2010), Kline (2011) and Byrne (2010) suggested that a model should satisfy the requirement of at least one index from each of the index categories, that is, absolute fit, incremental fit and parsimonious fit indices. In respect of the RMSEA,

Byrne (2010) pointed out that a range of 0.08 to 1.00 is also acceptable in assessing model fit. Hence from the figure 4 it can be concluded that the model has achieved indices criteria as in table 4

Table 4 - Structural Model's Achievement

Index Category	<b>Indices Used</b>	Acceptable level	Model's Achievement
Absolute fit	RMSEA	Value < 0.08	0.027
Absolute fit	GFI	Value > 0.90	0.907
Incremental fit	CFI	Value > 0.90	0.980
Parsimonious fit	Chi sq./df	Value < 5.0	1.225

Results from Table 4 indicates that all the indices of the final model have achieved the assessment criteria.

# 5. Hypothesis Testing

After completing the assessment of the measurement and structural mode, the following step is to check the model according to the proposed hypotheses. In the case of mediation model there are two types of hypotheses which are direct relationship and indirect relationship.

# 5.1 Direct Relationship

For the direct relationship, the path relationship/hypothesis as in Table 5 presented the standardized regression coefficients of the direct relationships of product innovation, market innovation, process innovation, organizational innovation and technology with the endogenous construct job performance.

Table 5 - Results of The Direct Relationship

Path relationship	Estimate	S.E	C.R.	P-value	$\mathbb{R}^2$
JP←Prod_Inno	.111	.107	1.201	.230	_
JP← Mkt_Inno	.125	.092	1.326	.185	-
JP← Proc_Inno	.084	.103	.889	.374	.84
JP←Org_Inno	.892	.064	5.517	***	-
JP←Techn	.016	.039	.174	.862	-
Techn←Prod_Inno	.062	.187	.877	.381	
Techn ←Mkt_Inno	004	.161	052	.959	.042
Techn ←Proc_Inno	074	.182	-1.027	.305	.042
Techn ←Org_Inno	180	.066	-2.482	.013	-

<sup>\*\*</sup>indicates significance at p<.05

From Table 5, it is shown that all the direct relationships reported a positive effect. However, only the path JP $\leftarrow$  Org\_Inno indicated a statistically significant effect ( $\beta$  = .892; CR= 5.517; p>.05) the remaining path coefficients are not statistically significant. Altogether, the four exogenous constructs together with the mediating construct explained 84% variation in job performance. Additionally, the table also presents the path relationship between the mediator construct, technology and the four exogenous constructs (product innovation, market innovation, process innovation, and organizational innovation). From the result, it is shown that collectively the four exogenous constructs explained less than 1% variability in technology.

#### 5.2 Indirect Relationship

To test the mediation effect of technology on the relationship between product innovation, market innovation, process innovation, organizational innovation and job satisfaction, the bootstrapping method is used. The bootstrapping method has been described as the most effective method of testing mediation against the Sobel Test method (Hayes, 2014). The procedure involved re-sampling of the working data set between 500 and 1000 times which a sampling distribution from which the total effect, the direct effect and indirect effect estimates and their corresponding 95% confidence interval values are produced. The algorithm also estimates the lower and upper limits as well as the two-tailed significant values for the effects. Table 6 shows the bootstrapping result for testing the mediation effect of technology in the research model.

**Table 6 - Results of Indirect Relationship** 

- *****						
Doth volotionship	β <b>value</b>	CI v	D volue			
Path relationship	Estimate	Lower Bounds	<b>Upper Bounds</b>	P-value		
JP←Techn←Prod_Inno	.001	013	.001	.523		
JP← Techn←Mkt_Inno	.000	.000	.000	.361		
JP← Techn←Proc_Inno	001	003	.002	.442		
JP←Techn←Org Inno	001	.002	.001	.497		

<sup>\*\*\*\*</sup>indicates significance at p<.05

Table 6 shows that technology does not show any mediation effect on the relationship between product innovation and job performance ( $\beta$ = .001; 95% CI: -.013~.001; p=.523). Similarly, job performance is shown to have no mediation effect on the relationship between market innovation and job performance ( $\beta$ = -.000; 95% CI: .000~.000; p=.361). In addition, technology does not mediate the relationship between process innovation, and organizational innovation and job performance with respective statistics as in the table.

#### 6. Conclusion

This study developed and assessed the mediation model using collected data from questionnaire survey amongst the staff in the Ministry of Culture, Youth and Community Development of UAE. It found that organizational innovation, product innovation, process innovation and marketing innovation has a negative relationship with technology. Moreover, organizational innovation, product innovation, process innovation and marketing innovation were positively associated with the employees' job performance. Unfortunately, this study was unable to establish the suggested positive impact of product innovation, process innovation and marketing innovation on technology maybe to the collected data were not strong enough to establish the relationship. Hopefully this study can be applicable to improve the job performance in the UAE and further explore by incorporating other variables to the mediation model

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