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# INFLUENCING FACTORS OF BUSINESS INTELLIGENCE CONTINUANCE USAGE INTENTION: A CASE STUDY OF GOVERNMENT-OWNED INSURANCE COMPANY

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Abstract— Business Intelligence is a technological trend in supporting strategic analysis and decision making in organizations. Many organizations claimed to have successfully adopted Business Intelligence, many also reported failure after adoption. Taspen is one of government-owned insurance company in Indonesia that has problems with the use of Business Intelligence, it decreased to reach 16% in the first year. Based on these facts it is necessary to analyze the factors that influence the continued use of Business Intelligence in the company. This study uses a quantitative approach through the distribution of questionnaires to decision makers and users of Business Intelligence. From the results of data collection, there were 195 respondents who were analyzed using Covariance Based SEM (CB-SEM) with the help of IBM Amos v22 software. This study proposes a framework model that combines External Stimuli, Affective Response, and Cognitive Response to the continuity of Business Intelligence use. The results of this study indicate that Information Integration and productivity increase are the most significant factors influencing the continuity of Business Intelligence use. Factors such as influence organizations and influence peers still need to be investigated further to identify the causes of the weak influence on the continued use of Business Intelligence.

# Keywords— Business Intelligence, CB-SEM, Continuance Usage Intention, Taspen

## I. INTRODUCTION

The growth of the current business environment requires organizations to provide a fast and comprehensive information flow. The ability to adapt to technological developments and manipulate large information is an important factor for improving decision-making processes and competing against competitors (Hannula and Virpi, 2003). There is a trend that organizations invest large costs in Business Intelligence to improve decision making capabilities. Based on Gartner's information technology investment survey, Business Intelligence technology is in the top priority of Chief Information Officers (CIO). When IT investment only increased by 0.4%, investment in Business Intelligence technology increased by 7.9%.

Although the development and investment of Business Intelligence is growing rapidly, many organizations have complained of Business Intelligence that is not in accordance with the expected results. A survey conducted by 200 decision makers showed that the majority of organizations have spent billions of Rupiah on Business Intelligence, but 62% of respondents stated that users did not get the expected results. According to Orlikowski and Iacono (2001), there is no single conceptual snapshot of IT that will capture all usage contexts. Unlike other operational systems, Business Intelligence is a challenge in itself because it is a large enterprise system, longterm maintenance, and difficult to assess. (Seddon et al, 2010).

To understand user behavior, it is very important to consider the continued use of an information system when the user has been using information system for a long time in daily activities (Bhattacherjee, 2001). Realizing the need to better understand continued IS usage behavior, researchers have recently begun to study the subject in more detail (Karahanna et al, 1999; Bhattacherjee 2001; Venkatesh 2002). The focus of research on IT adoption and post-adoption or IS continuance has been on mainly cognition-oriented behavior models, such as the technology acceptance model (TAM) (Davis, 1989), the Theory of Planned Behavior (Ajzen, 2001), and the IS continuance model (Bhattacherjee, 2001).

Taspen is a company engaged in the field of insurance for civil servants. Taspen has Business Intelligence as one of the Business Management Programs in the Information Technology Master Plan (ITMP). Based on data from Business Intelligence managers, it is known that the use of Business Intelligence does not reach the target. Until the end

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of 2019, active users of Business Intelligence do not exceed 16%. The gap between reality and expectations raises the problem of the successful use of Business Intelligence. According to Rasanen (2001), business success is dependent upon three factors. A company must be able (1) to satisfy its core business agenda and, secondly (2) to form product entity that is successful among the customers. The third is (3) the external business environment. Management needs to predict any changes and make decisions based upon flawless real-time information. This is so because, given its critical role in today's business processes, infrequent, inappropriate, and ineffective long-term use of IS often contributes to corporate failures (Lyytinen and Hirschheim, 1987). Continuance is not entirely an alien concept in IS research. It has been examined variously as "implementation" (Zmud 1982), "incorporation" (Kwon and Zmud 1987), and "routinization" (Cooper and Zmud 1990) in the IS implementation literatur.

Based on this, it is necessary to analyze the factors that influence the continued use of Business Intelligence in the company. This study proposes a model to identify External, Internal, Affective and Cognitive factors, which consists of motivational factors, success factors, beneficial factors, and organizational factors towards the survival of Business Intelligence. These problems will be analyzed further in the research.

### II. **DESIGN OVERVIEW**

The method used in this research is a case study. The study was conducted using a quantitative approach through the distribution of questionnaires to Business Intelligence users. The questionnaire statement was divided into two parts, the first was related to respondent demographics, and the second was related to research statements. The first part contains the demographics of respondents consisting of gender, age, education, position, work unit, experience and frequency using Business Intelligence, the next section is a statement that represents the variables in the research model. The statement uses a likert scale with an interval of 1 to 6. Table 1 Dessenab design

Element	Information				
Classification	Case Study				
Paradigm	Evaluative				
Research purposes	To identify the factors that influence the use of Business Intelligence				
Data Type	Likert scale with interval 1 to 6.				
Data Collection	Online and offline questionnaire				
Data Analysis	Quantitative				
Method of Conclusions	Deductive				
Data Processing Method	SEM (Structural Equation				
	Modeling)				
Research Instrument	6 independent variables and 6 dependent variables				

Data Source Population	195 respondent (error level 5%)				
Data processing tools	Microsoft Excel, SPSS, dan Amos				

This study uses non-probability sampling where research subjects are not taken randomly but rather determined by the authors. The author uses a purposive sampling method where users have used Business Intelligence for a certain period. Data collection is carried out for approximately one month through online and offline media. The population of Business Intelligence users is 250 users, consisting of the board of directors, General Manager, Head of Work Unit, Manager, Assistant Manager, functional employees, and staff. From the results of data collection, 195 respondents were obtained with an error level of 5%, then an analysis was performed using Covariance Based SEM (CB-SEM) with the help of the IBM Amos v22 application. Data analysis with SEM aims to determine the level of validity of the construct and to determine the level of reliability of the indicators used, as well as the level of significance of the relationships between constructs.

### III. EXPERIMENT AND RESULT

Based on the data, it can be seen that respondents are dominated by men with a percentage of 70.3%, respondents aged between 21-29 years with a percentage of 47%, for more details see Table 2.

Variable	Category	Percent
Sor	Man	70,3%
Sex	Woman	29,7%
	21-29 years old	47%
A	> 50 years old	30%
Age	30-39 years old	13%
	40-49 years old	9%
	Master	14,9%
Education	Bachelor	61,5%
Education	Diploma	19%
	High School	4,6%
	Staff	49,74%
	Functional Staff	22,56%
Desition	Assistant Manager	5,64%
rosition	Manager	15,38%
	General Manager	6,15%
	Board of Director	0,51%
Experience of Using	Less than 1 year	61,5%
Business Intelligence	1-2 years	24,6%

Table 2. Respondent demography

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	> 2 years	13,8%
Frequency of Using	More than 1 month	34,4%
	At least once a month	28,7%
Business Intelligence	At least once a week	17,4%
	Almost Everyday	13,5%

The author conducts the measurement model test and the structural model test. The measurement model test aims to find out how precisely the indicator explains its latent construct. while the structural model test determines whether a hypothesis is rejected or accepted. The measurement model test consists of the validity test, the reliability test, the goodness of fit test, while the structural model test consists of the hypothesis test and the significance of the relationship between constructs. Validity test is divided into convergent validity test to see the loading factor value of each indicator, and to see the average variance extracted (AVE) value. The discriminant validity test compares the loading indicator value with cross loading, and compares the AVE square root value of a construct with the value of correlation between constructs. Reliability test is performed to see the value of construct reliability (CR) and cronbach's alpha (CA) of each construct. The CR and CA values for each construct must be  $\geq 0.7$ .

 Table 3. Measurement model test

Indicator	Estimate	AVE	CA	CR	
II3	0.897		0.874936	0.895	
II2	0.845	0.700777			
II1	0.764				
IP1	0.911		0.941661	0.957	
IP2	0.913	0 201/05			
IP3	0.897	0.001495			
IP4	0.859				
AR3	0.908		0.879653	0.905	
AR2	0.791	0.709707			
AR1	0.824				
C1	0.924		0.913218	0.913	
C2	0.845	0.778392			
C3	0.876				
TR3	0.698		0.885429	0.896	
TR2	0.89	0.65621			
TR1	0.944				
IQ3	0.768		0.850995	0.86	
IQ2	0.866	0.65621			
IQ1	0.793				
CU3	0.923	0.688705	0.867898	0.89	

CU2	0.719			
CU1	0.835			
USE1	0.819			
USE2	0.876	0.749979	0.899849	0.923
USE3	0.901			
RE3	0.852			
RE2	0.945	0.794851	0.920639	0.919
RE1	0.875			
S3	0.914			
S2	0.86	0.66068	0.851082	0.847
<b>S</b> 1	0.638			
IFP3	0.932			
IFP2	0.88	0.823708	0.933375	0.933
IFP1	0.91			
IO3	0.854			
IO2	0.885	0 655907	0 992164	0.870
IO1	0.693	0.055607	0.003104	0.079
IO4	0.794			

The validity test on the measurement model test shows that the comparison of loading indicator values is greater than cross loading, the comparison of AVE squared values is greater than the correlation between constructs. These results indicate that all indicators and variables in the test measurement model are valid and reliable.

Structural model test consists of hypothesis test and relationship between constructs test. The hypothesis is accepted if the p-value  $< \alpha$  (0.05) and | C.R. | > 1.96. For relationship between constructs test, if the estimated value > 0.5, then both constructs have a strong relationship, whereas if the estimate value <0.5 then the two constructs have a weak relationship.

Table 4. Goodness of fit test results

Criteria	Cut off value	GOF	Result			
		Score				
CMIN/DF	$\leq 2 \operatorname{atau} \leq 3$	1.315	Good fit			
GFI	> 0,8	0.842	Good fit			
RMR	As low as	0.042	Good fit			
	possible					
RMSEA	$\leq$ 0,05	0.043	Good fit			
TLI	$\geq 0,9$	0.967	Good fit			
NFI	$\geq 0,9$	0.906	Good fit			
CFI	$\geq 0,9$	0.975	Good fit			

Table 5. Structural Model Test Results

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Varia	ble		C.R.	Р	Test Results	Estimate
InProductivity	÷	InfInteg ration	7.122	***	Accepted	0.736
AdReporting	4	InfInteg ration	6.108	***	Accepted	1.449
Customization	÷	InfInteg ration	3.13	0	Accepted	0.424
InProductivity	÷	InfQual ity	1.873	0.06	Rejected	0.23
AdReporting	÷	InfQual ity	-2.224	0.03	Accepted	-0.62
Customization	÷	InfQual ity	2.411	0.02	Accepted	0.419
Trust	÷	InProdu ctivity	3.318	***	Accepted	0.254
Satisfaction	+	InProdu ctivity	6.327	***	Accepted	0.863
Trust	÷	AdRep orting	-2.923	0	Accepted	0.714
Satisfaction	4	AdRep orting	7.235	***	Accepted	-0.111
Trust	÷	Custom ization	-0.953	0.34	Rejected	-0.26
Satisfaction	¢	Custom ization	0.384	0.7	Rejected	0.034
BIConUse	¢	Trust	0.352	0.73	Rejected	0.241
BIConUse	4	Satisfac tion	2.06	0.04	Accepted	0.03
BIConUse	¢	InfPeer s	0.382	0.7	Rejected	0.037
BIConUse	¢	InfOrga nization s	1.457	0.15	Rejected	0.132
BIConUse	÷	Compet ence	0.404	0.69	Rejected	0.031
BIConUse	÷	Related	3.834	***	Accepted	0.371

The results of the structural test show that not all hypotheses put forward by the author are accepted, there are seven hypotheses rejected. However, indicators and variables in the structural model test are valid and reliable. Based on the results of the model test, a concept of model that affects the continued use of Business Intelligence is formed.



Fig 1. Research model

### IV. CONCLUSION

Based on research findings, it can be concluded that:

- 1. The most significant factor from the external side that affects the continued use of Business Intelligence is information Integration with a value of |C.R| = 7.122 and p-value = \*\*\*. This shows that users will continue to use Business Intelligence because it is able to integrate information from various sources to increase user productivity.
- The most significant factor in influencing the continued 2. use of Business Intelligence in terms of the perceive of usefulness is increase productivity with a value of | C.R | = 7,235 and p-value = \*\*\*. Users will continue to use BI because it can improve performance productivity, assist in better decision making, analyze complex problems and produce reports quickly.
- In terms of internal stimuli, although not significant, the relatedness factor has an influence on the continuance use of Business Intelligence, with a value of |C.R| = 3,834and p-value = \*\*\*. This indicates that users feel valued by the organization, and feel accepted by the environment, thus influencing their desire to continue to use Business Intelligence.
- 4. Although not significant, the trust factor is a cognitive response that has an influence on the continued use of Business Intelligence, with a value of |C.R| = 2.06 and pvalue = 0.04. The influence of trust has not significantly affected the continued use of Business Intelligence. There are still issues of trust in the validity of data that have implications for decision making using Business Intelligence.
- influence 5. Organizational factors consisting of organizations and influence peers do not have an

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influence on the continuance use of Business Intelligence. The influence factor of organization has a value of | C.R | = 1,457 and p-value = 0.15, and the influence peers factor has a value of | C.R | = 0.382 and p-value = 0.7. This indicates that there is no management support in using Business Intelligence and the organization has not tried to increase interaction between Business Intelligence users.

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