

Chapter 11

Application of the Model: Case-Based Studies

ABSTRACT

It is argued that models are conceptualized, designed, developed, and validated to understand complex behaviour of larger entities. Models provide indicative measurements, and their validations in real life situation need careful considerations of relevant ambient conditions. Models also provide suggestive and causal relationships among their qualitative and quantitative influencers for better predictability. Generally, predictive models provide structural equations, measurement equations with associated random errors. These errors do play vital roles in relating abstracted behavior of the model outputs with the real life situations. In order to reduce these errors to an agreed level, case-based validations of models are quite important. This chapter discusses derived measurement and structural equations that the model has produced and presents some cases to examine the appropriateness of the application of the model developed.

DERIVED MEASUREMENT EQUATIONS OF THE MODEL

Measurement equations show the relationships among latent and exogenous variables. “Classical test theory” is the basis for establishing this relationship (Cronbach et al., 1972; Linn and Werts, 1979; Susan et al., 2008; España et al., 2010). The general form of the measurement equation is shown in Figure 1.

The measurement equation here is $X = \xi + \delta$ where X is the exogenous variables which is the mean of the items summated through dummy

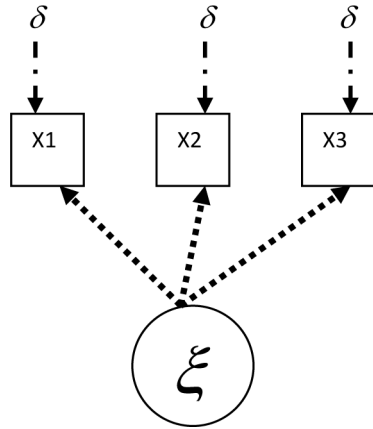
coding as explained in chapters eight and nine. ξ represents the endogenous variable in the equation. Each variable X thus represents a set of independent indicators (questions) administered to the respondents and ordinal data are acquired. Classical test theory considers measurement errors (δ) and provides a predictive ability to the relationship between exogenous variable and predictors. Table 1 describes the relationships and displays the foundation for such predictions.

This table provides set of questions for each of the variables. Its values are obtained from the survey instruments administered among the respondents. Using these data the values of the predicted latent exogenous variables were com-

DOI: 10.4018/978-1-4666-4201-0.ch011

Application of the Model

Figure 1. Measurement equation (SEM) (España et al., 2010)



puted through measurement equations shown in the Table 1. The variable “X” with its measurement errors is obtained for each respondent for each variable and its mean across the variables in each category is computed for analysis. Table 1 indicates that latent exogenous variables U, I, T compute latent endogenous variables (ξ) P. Similarly, C1, C2, C3 variables compute C and VC, AT variables compute AQ. These variables are computed based on the summative nature of the structural equation model (Pedhazur, 1997). For example, in order to compute pre-acquisition preparedness (P), P_U, P_I, P_T are computed for the

Table 1. Measurement equation

Stratified Independent Indicators Questions	Dummy Codes Used Item	Summated Independent Indicator Variable X	Description of Exogenous Variable X	Latent Endogenous Variables ξ	Measurement Equation (Classical Test Theory) ($X = \xi + \delta$)
U101-U106	U1	U	User Preparedness	P Pre-Acquisition Preparedness	$U = P_U + \delta$ $I = P_I + \delta$ $T = P_T + \delta$ $(P_{mean}) = \frac{P_U + P_I + P_T}{3}$
U201-U207	U2				
U301-U307	U3				
I101-I106	I1	I	IS Preparedness		
I201-I204	I2				
I301-I306	I3				
T101-T107	T1	T	Technology Preparedness		
T201-T209	T2				
T301-T307	T3				
C101-C108	C1	C1	User Perception on Organization	C Climate Preparedness	$C1 = C_{c1} + \delta$ $C2 = C_{c2} + \delta$ $C3 = C_{c3} + \delta$ $(C_{mean}) = \frac{C_{c1} + C_{c2} + C_{c3}}{3}$
C201-C207	C2	C2	User Perception on IT		
C301-C307	C3	C3	User Perception on Decision Making Process		
VC01-VC06	VC	VC	Vendor Management	AQ Acquisition Process	$VC = AQ + \delta$ $AT = AQ + \delta$ $(AQ_{mean}) = \frac{AQ_{vc} + AQ_{at}}{2}$
AT01-A06	AT	AT	Acquisition Process		

23 more pages are available in the full version of this document, which may be purchased using the "Add to Cart" button on the publisher's webpage:
www.igi-global.com/chapter/application-model-case-based-studies/76981

Related Content

An Intelligence-Based Model for Condition Monitoring Using Artificial Neural Networks

K. Jenab, K. Rashidiand S. Moslehpour (2013). *International Journal of Enterprise Information Systems* (pp. 43-62).

www.irma-international.org/article/an-intelligence-based-model-for-condition-monitoring-using-artificial-neural-networks/100382

Selfish Users and Fair Sharing of Bandwidth in Distributed Medium Access

Ratan K. Guhaand Sudipta Rakshit (2006). *International Journal of Enterprise Information Systems* (pp. 28-44).

www.irma-international.org/article/selfish-users-fair-sharing-bandwidth/2100

Human Resource Information System Use, Satisfaction, and Success

Sonalee Srivastava, Santosh Devand Badri Bajaj (2021). *International Journal of Enterprise Information Systems* (pp. 106-124).

www.irma-international.org/article/human-resource-information-system-use-satisfaction-and-success/268365

Implementation of Integrated Enterprise Asset Management Systems (IEAMS): Key Challenges and Lessons Learned

Asim Hussain (2013). *Cases on Enterprise Information Systems and Implementation Stages: Learning from the Gulf Region* (pp. 215-239).

www.irma-international.org/chapter/implementation-integrated-enterprise-asset-management/70311

Conceptual Foundations of Digital Twin-Enabled Smart and Resilient Homes

Manuel Villasaleroand Kalin Dimtchev (2026). *Digital Twin Applications and Cognitive Enterprise Transformation Across Industries* (pp. 83-106).

www.irma-international.org/chapter/conceptual-foundations-of-digital-twin-enabled-smart-and-resilient-homes/410285