

Social Media Marketing Management

社會媒體行銷管理

確認性因素分析

(Confirmatory Factor Analysis)

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Tue 12,13,14 (19:20-22:10) D325

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課程大綱 (Syllabus)

週次	日期	內容 (Subject/Topics)
1	102/02/19	社會媒體行銷管理課程介紹 (Course Orientation of Social Media Marketing Management)
2	102/02/26	社群網路 (Social Media: Facebook, Youtube, Blog, Microblog)
3	102/03/05	社群網路行銷 (Social Media Marketing)
4	102/03/12	行銷管理 (Marketing Management)
5	102/03/19	社群網路服務與資訊系統理論 (Theories of Social Media Services and Information Systems)
6	102/03/26	行銷理論 (Marketing Theories)
7	102/04/02	教學行政觀摩日 (Off-campus study)
8	102/04/09	行銷管理論文研討 (Paper Reading on Marketing Management)
9	102/04/16	社群網路行為研究 (Behavior Research on Social Media)

課程大綱 (Syllabus)

週次	日期	內容 (Subject/Topics)
10	102/04/23	期中報告 (Midterm Presentation)
11	102/04/30	社群網路商業模式 [Invited Speaker: Dr. Rick Cheng-Yu Lu] (Business Models and Issues of Social Media)
12	102/05/07	社群網路策略 (Strategy of Social Media)
13	102/05/14	社群口碑與社群網路探勘 (Social Word-of-Mouth and Web Mining on Social Media)
14	102/05/21	社群網路論文研討 (Paper Reading on Social Media)
15	102/05/28	探索性因素分析 (Exploratory Factor Analysis)
16	102/06/04 (> 6/01)	確認性因素分析 (Confirmatory Factor Analysis)
17	102/06/11 (> 6/04)	期末報告1 (Term Project Presentation 1)
18	102/06/18 (> 6/11)	期末報告2 (Term Project Presentation 2)

Types of Factor Analysis

- **Exploratory Factor Analysis (EFA)**
 - is used to discover the factor structure of a construct and examine its reliability.
It is **data driven**.
- **Confirmatory Factor Analysis (CFA)**
 - is used to confirm the fit of the hypothesized factor structure to the observed (sample) data.
It is **theory driven**.

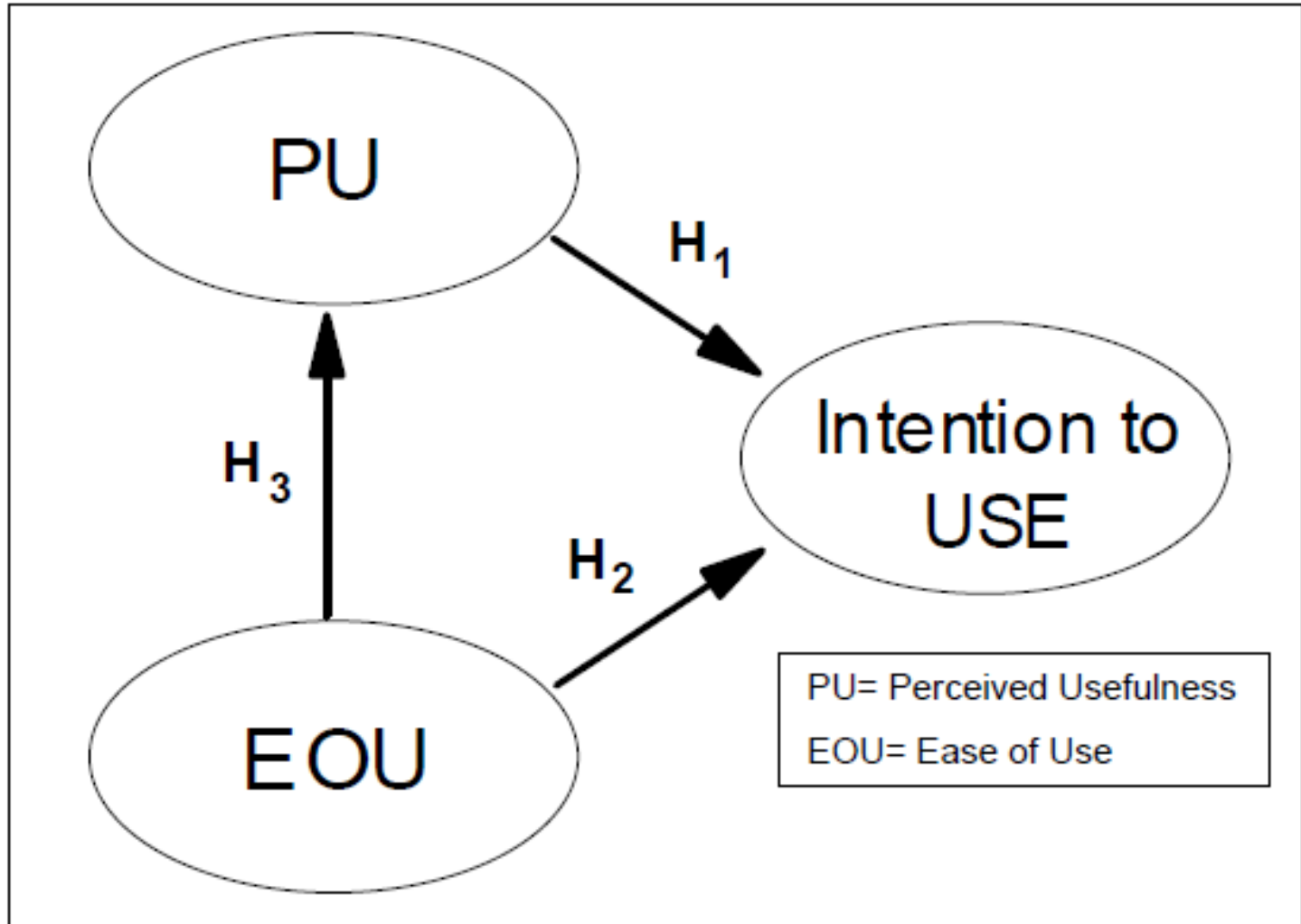
Structural Equation Modeling (SEM)

- Structural Equation Modeling (SEM)
techniques such as
LISREL and
Partial Least Squares (PLS)
are
second generation data analysis techniques

Data Analysis Techniques

- **Second** generation data analysis techniques
 - SEM
 - PLS, LISREL
 - statistical conclusion validity
- **First** generation statistical tools
 - Regression models:
 - linear regression, LOGIT, ANOVA, and MANOVA

The TAM Model



Structured Equation Modeling (SEM)

- Structural model
 - the assumed causation among a set of **dependent** and **independent** constructs
- Measurement model
 - **loadings** of **observed items (measurements)** on their **expected latent variables (constructs)**.

Structured Equation Modeling (SEM)

- The combined analysis of the **measurement** and the **structural** model enables:
 - measurement errors of the **observed variables** to be analyzed as an integral part of the model
 - **factor analysis** to be combined in one operation with the **hypotheses testing**
- SEM
 - **factor analysis** and **hypotheses** are tested in the same analysis

Use of Structural Equation Modeling Tools 1994-1997

SEM Approaches	I&M (n=106)	ISR (n=27)	MISQ (n=38)	All Three Journals
PLS	2%	19%	11%	7%
LISREL	3%	15%	11%	7%
Other *	3%	11%	3%	4%
Total %	8%	45%	25%	18%

* Other includes SEM techniques such as [AMOS](#) and [EQS](#).

SEM models in the IT literature

- Partial-least-squares-based SEM
 - PLS
- Covariance-based SEM
 - LISREL

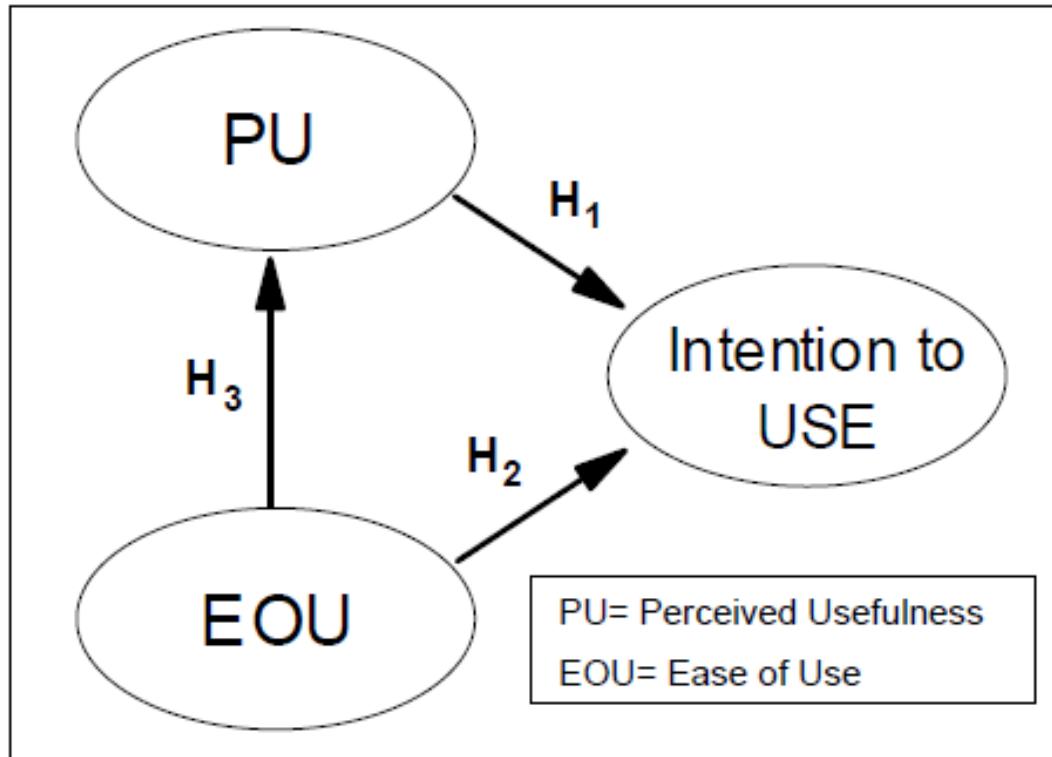
Comparative Analysis between Techniques

Issue	LISREL	PLS	Linear Regression
Objective of Overall Analysis	Show that the null hypothesis of the entire proposed model is plausible, while rejecting path-specific null hypotheses of no effect.	Reject a set of path-specific null hypotheses of no effect.	Reject a set of path-specific null hypotheses of no effect.
Objective of Variance Analysis	Overall model fit, such as insignificant χ^2 or high AGFI.	Variance explanation (high R-square)	Variance explanation (high R-square)
Required Theory Base	Requires sound theory base. Supports confirmatory research.	Does not necessarily require sound theory base. Supports both exploratory and confirmatory research.	Does not necessarily require sound theory base. Supports both exploratory and confirmatory research.
Assumed Distribution	Multivariate normal, if estimation is through ML. Deviations from multivariate normal are supported with other estimation techniques.	Relatively robust to deviations from a multivariate distribution.	Relatively robust to deviations from a multivariate distribution, with established methods of handling non-multivariate distributions.
Required Minimal Sample Size	At least 100-150 cases.	At least 10 times the number of items in the most complex construct.	Supports smaller sample sizes, although a sample of at least 30 is required.

Capabilities by Research Approach

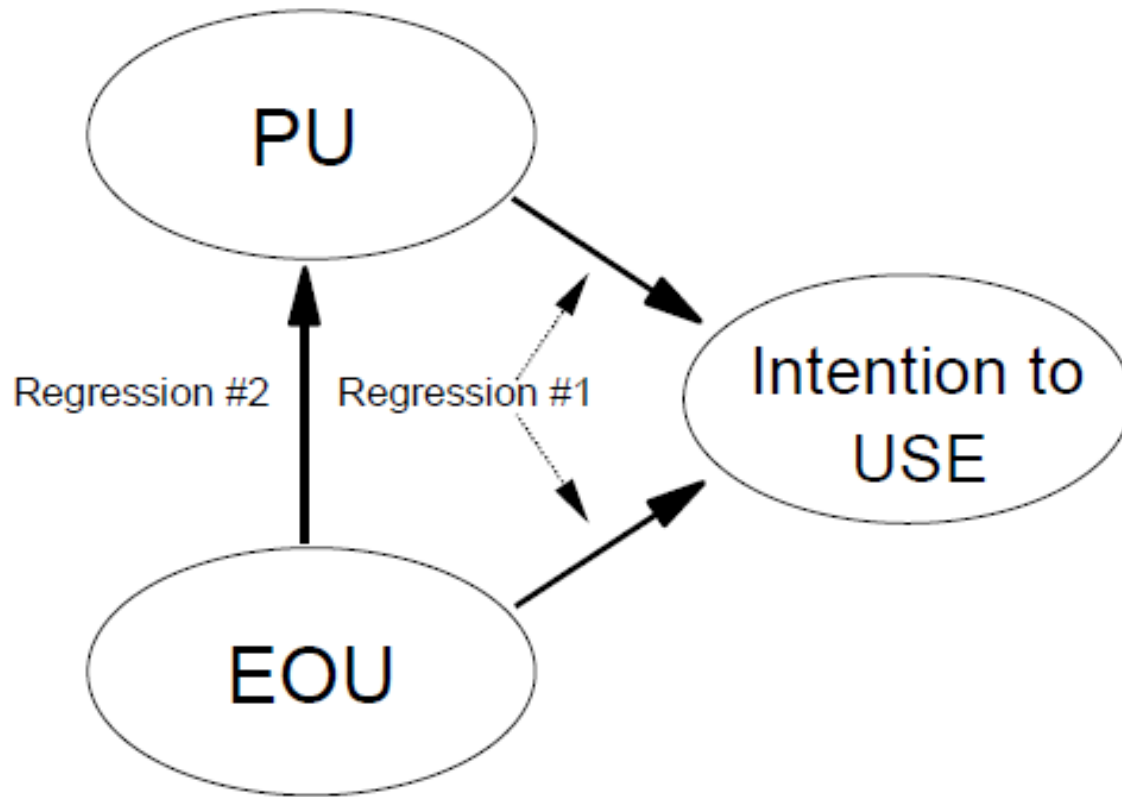
Capabilities	LISREL	PLS	Regression
Maps paths to many dependent (latent or observed) variables in the same research model and analyze all the paths simultaneously rather than one at a time.	Supported	Supported	Not supported
Maps specific and error variance of the observed variables into the research model.	Supported	Not supported	Not supported
Maps reflective observed variables	Supported	Supported	Supported
Maps formative observed variables	Not supported	Supported	Not supported
Permits rigorous analysis of all the variance components of each observed variable (common, specific, and error) as an integral part of assessing the structural model .	Supported	Not supported	Not supported
Allows setting of non-common variance of an observed variable to a given value in the research model.	Supported	Not supported	Supported by adjusting the correlation matrix.
Analyzes all the paths, both measurement and structural, in one analysis.	Supported	Supported	Not supported
Can perform a confirmatory factor analysis	Supported	Supported	Not supported
Provides a statistic to compare alternative confirmatory factor analyses models	Supported	Not supported	Not supported

TAM Model and Hypothesis



	Hypothesis
H ₁	<u>PU</u> will impact the system outcome construct, Intention to Use the System.
H ₂	<u>EOU</u> will impact the system outcome construct, Intention to Use the System.
H ₃	<u>EOU</u> will impact <u>PU</u> .

TAM Causal Path Findings via Linear Regression Analysis



	DV	F (R ²)	IV	Coefficient (T-value)
Regression #1	Intention to Use	23.80** (.24)	PU	.41 (4.45**)
			EOU	.10 (1.07)
Regression #2	PU	124.01** (.44)	EOU	.66 (11.14**)

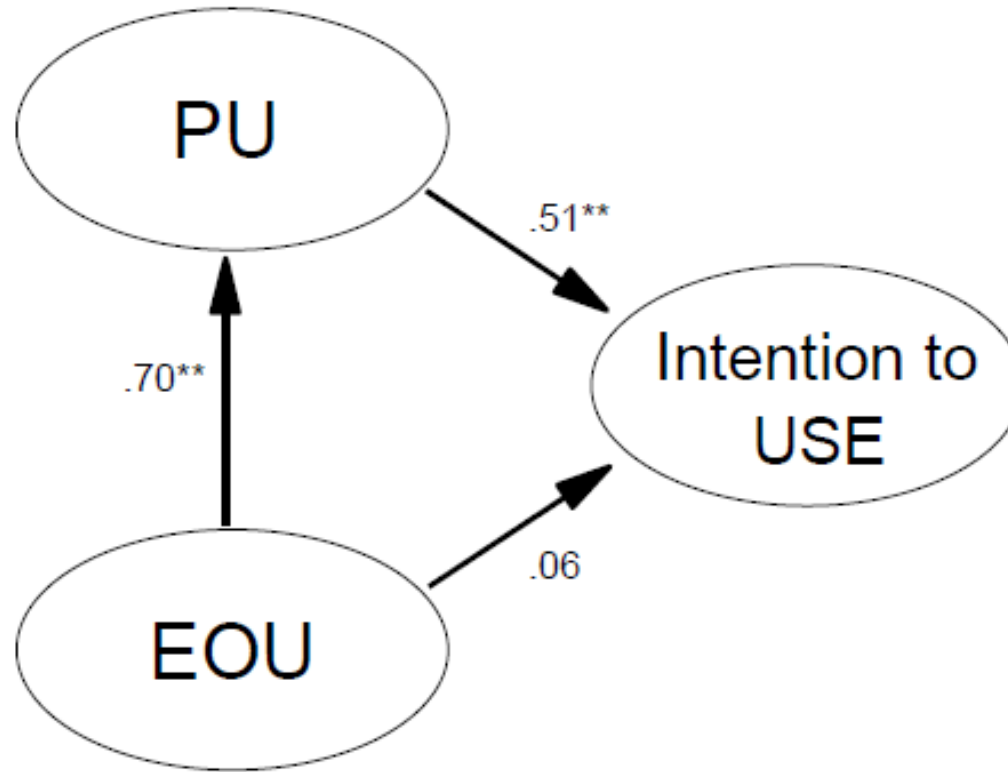
** = Significant at the .01 level

Factor Analysis and Reliabilities for Example Dataset

Construct	Item	Factors			Cronbach's α
		1	2	3	
Perceived Usefulness (PU)	PU1	.543	.277	.185	.91
	PU2	.771	.178	.053	
	PU3	.827	.315	.185	
	PU4	.800	.268	.234	
	PU5	.762	.352	.236	
	PU6	.844	.437	.290	
Perceived Ease-of-Use (EOU)	EOU1	.265	.751	.109	.93
	EOU2	.217	.774	.150	
	EOU3	.270	.853	.103	
	EOU4	.303	.787	.105	
	EOU5	.248	.831	.179	
	EOU6	.242	.859	.152	
Intention To Use (IUSE)	IUSE1	.183	.147	.849	.80
	IUSE2	.224	.062	.835	
	IUSE3	.139	.226	.754	

Rotation Method: Varimax with Kaiser Normalization (Rotation converged in 6 iterations)

TAM Standardized Causal Path Findings via LISREL Analysis



LISREL Fit Indices
$X^2 = 160.17$
df = 87
AGFI = .84
RMR = .047

Link	Coefficient (T-value)	SMC
PU -> Intended Use	.51 (3.94**)	.30
EOU -> Intended Use	.06 (.48)	
EOU -> PU	.70 (7.05**)	.48

** = Significant at the .01 level

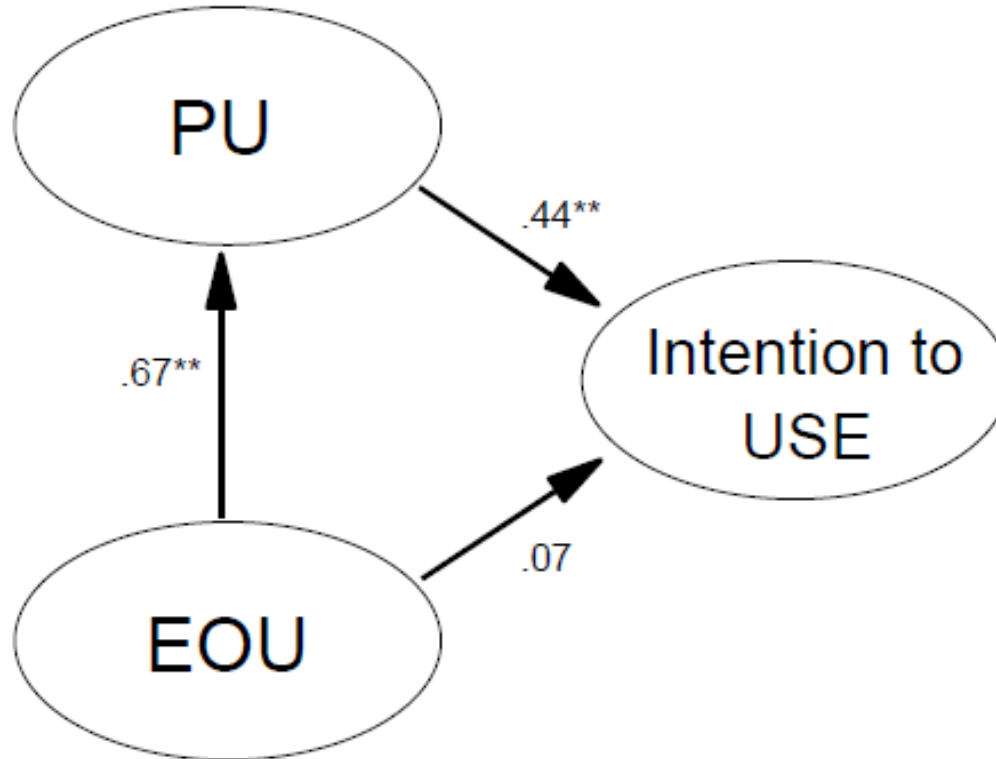
Standardized Loadings and Reliabilities in LISREL Analysis

Construct	Item	Latent Construct Loading (and Error)			Reliability Coefficient
		<i>PU</i>	<i>EOU</i>	<i>IUSE</i>	
Perceived Usefulness (PU)	PU1	0.99 (.50)			.95
	PU2	1.10 (.39)**			
	PU3	0.93 (.45)**			
	PU4	1.07 (.26)**			
	PU5	1.10 (.29)**			
	PU6	1.11 (.24)**			
Perceived Ease-of-Use (EOU)	EOU1		0.78 (.45)		.94
	EOU2		0.95 (.38)**		
	EOU3		0.92 (.25)**		
	EOU4		0.99 (.31)**		
	EOU5		1.00 (.27)**		
	EOU6		0.94 (.21)**		
Intention To Use (IUSE)	IUSE1			1.36 (.34)	.95
	IUSE2			2.17 (.38)**	
	IUSE3			1.15 (.53)**	

The first item [loading](#) in each [latent variable](#) is fixed at 1.00 and does not have a t- value.

** Significant at the .01 level

TAM Causal Path Findings via PLS Analysis



Link	Coefficient (T-value)	R ²
PU → Intended Use	.44 (3.69**)	.24
EOU → Intended Use	.07 (.12)	
EOU → PU	.67 (10.20**)	.44

** = Significant at the .01 level

Loadings in PLS Analysis

Construct	Item	Latent Construct		
		<i>PU</i>	<i>EOU</i>	<i>IUSE</i>
Perceived Usefulness (PU)	PU1	.776**	.613	.405
	PU2	.828**	.498	.407
	PU3	.789**	.448	.302
	PU4	.886**	.558	.353
	PU5	.862**	.591	.451
	PU6	.879**	.562	.406
Perceived Ease-of-Use (EOU)	EOU1	.534	.802**	.323
	EOU2	.557	.839**	.338
	EOU3	.467	.886**	.260
	EOU4	.562	.843**	.289
	EOU5	.542	.865**	.304
	EOU6	.508	.889**	.288
Intention To Use (IUSE)	IUSE1	.350	.270	.868**
	IUSE2	.380	.234	.858**
	IUSE3	.336	.280	.814**

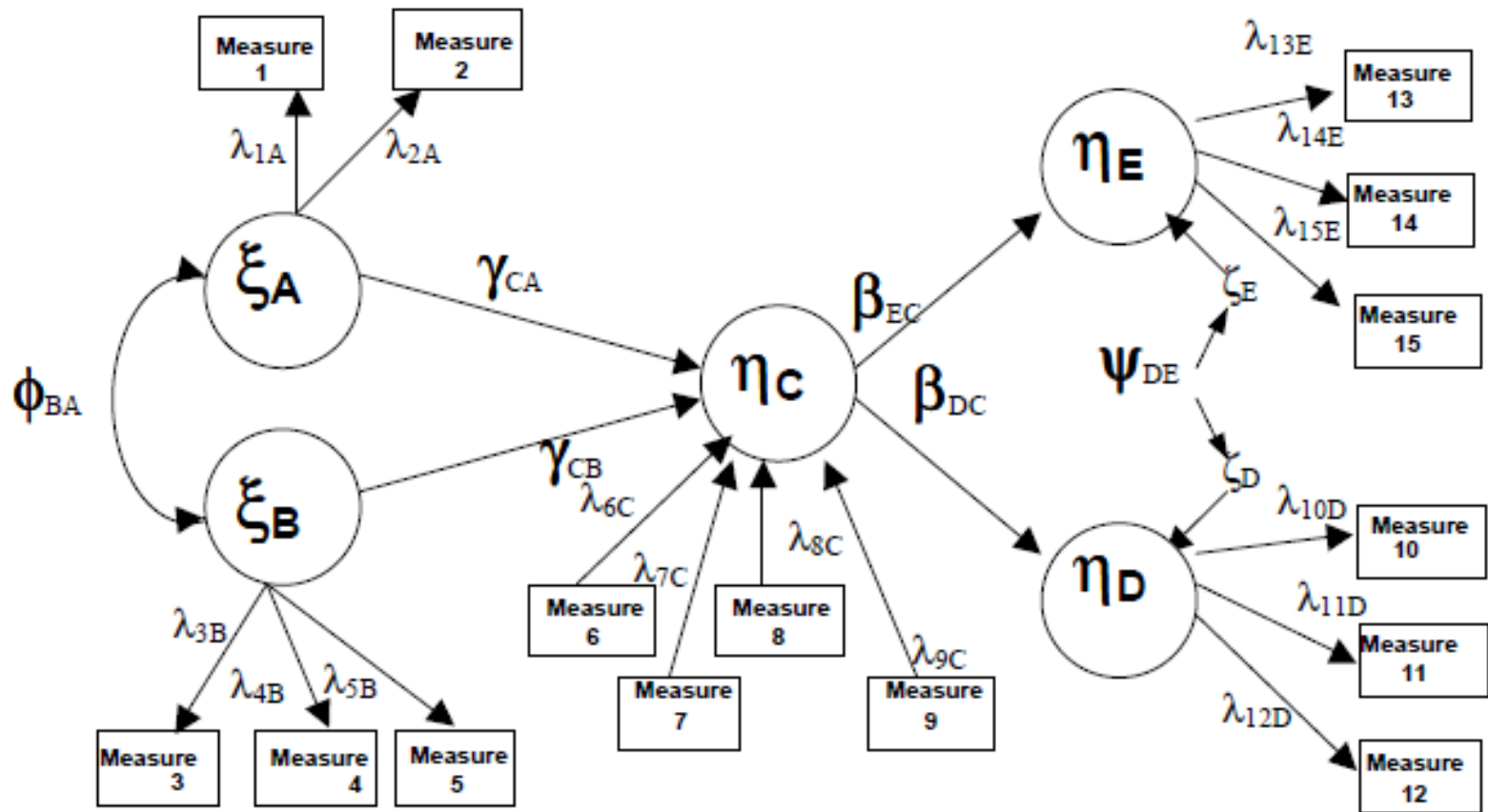
N.B. A reliability statistic not automatically produced in PLS.

** Significant at the .01 level

AVE and Correlation Among Constructs in PLS Analysis

AVE/ Correlation	IUSE	PU	EOU
IUSE	.721		
PU	.468	.742	
EOU	.359	.632	.738

Generic Theoretical Network with Constructs and Measures



Exogenous Latent Variables A and B

Endogenous Latent Variables C, D, and E

Number Of Covariance-based SEM Articles Reporting SEM Statistics in IS Research

Statistics	<i>I&M</i> (n=6)	<i>ISR</i> (n=7)	<i>MISQ</i> (n=5)	All Journals (n=18)
GFI reported	3 (50%)	3 (43%)	1 (20%)	7 (39%)
Of GFI reported, number > 0.90	1 (33%)	2 (67%)	1 (100%)	4 (57%)
AGFI reported	2 (33%)	2 (29%)	1 (20%)	5 (28%)
Of AGFI reported, number > 0.80	1 (50%)	2 (100%)	1 (100%)	4 (80%)
RMR reported	2 (33%)	4 (57%)	2 (40%)	8 (44%)
Of RMR reported, number < 0.05	0 (0%)	1 (25%)	1 (50%)	2 (25%)
χ^2 insignificance reported	3 (50%)	2 (29%)	0 (0%)	5 (28%)
Of χ^2 insig. reported, number > .05	3 (100%)	1 (50%)	0 (0%)	4 (80%)
Ratio χ^2 / df reported	5 (83%)	6 (86%)	4 (80%)	15 (83%)
Of ratio χ^2 / df reported, number < 3	5 (100%)	5 (83%)	2 (50%)	12 (80%)
SMC	2 (33%)	3 (43%)	2 (40%)	7 (39%)
NFI reported	3 (50%)	3 (43%)	3 (60%)	9 (50%)
Of NFI reported, number > .90	2 (67%)	3 (100%)	3 (100%)	8 (89%)
CFI reported	3 (50%)	2 (29%)	1 (20%)	6 (33%)
T-values or significance of paths	4 (67%)	6 (86%)	4 (80%)	14 (78%)
Construct Reliability reported	5 (83%)	7 (100%)	4 (80%)	16 (89%)
Use of Nested Models	4 (67%)	6 (86%)	3 (60%)	13 (72%)

Notes: Rows in gray should receive special attention when reporting results
 11 articles used LISREL, 6 [EQS](#), and 1 [AMOS](#)

Number of PLS Studies Reporting PLS Statistics in IS Research (Rows in gray should receive special attention when reporting results)

PLS Statistics	<i>I&M</i> (n=2)	<i>ISR</i> (n=5)	<i>MISQ</i> (n=4)	All Journals (n=11)
<u>R²</u> reported	2 (100%)	5 (100%)	4 (100%)	11 (100%)
<u>AVE</u> reported	2 (100%)	5 (100%)	3 (75%)	10 (91%)
T-values or significance of paths	2 (100%)	5 (100%)	4 (100%)	11 (100%)
Construct Reliability reported	2 (100%)	4 (80%)	3 (75%)	9 (82%)
Use of <u>Nested Models</u>	0 (0%)	0 (0%)	0 (0%)	0 (0%)

In LISREL terminology, the structural model contains the following:

- exogenous latent constructs called Xi or Ksi (ξ), depending on the dictionary used.
- endogenous latent constructs called Eta (η).
- paths connecting ξ to η represented statistically as Gamma (γ) coefficients.
- paths connecting one η to another are designated Beta (β).
- shared correlation matrix among ξ ; called Phi (ϕ).
- shared correlation matrix among the error terms of the η called Psi (ψ).
- the error terms themselves are known as ζ (Zeta).

To illustrate, IUSE and PU would be considered to be endogenous constructs in the TAM running example used earlier. Both are predicted by one or more other variables, or latent constructs. EOU, however, would be considered to be an exogenous latent construct in that no other variable in this particular model predicts it. The causal path PU (ξ_1) \Rightarrow IUSE (ξ_2) was estimated as a β coefficient. The causal path EOU (η_1) \Rightarrow PU (ξ_1) was estimated as a γ coefficient.⁹

In addition, the [measurement model](#) consists of:

- X and Y variables, which are observations or the actual data collected. X and Y are the measures of the [exogenous](#) and [endogenous](#) constructs, respectively. Each X should load onto one ξ , and each Y should load onto one η .
- Lambda X (λ_X) representing the path between an observed variable X and its ξ , i.e., the item [loading](#) on its [latent variable](#).
- Theta Delta (Θ_δ) representing the error variance associated with this X item, i.e., the variance not reflecting its [latent variable](#) ξ .
- Lambda Y (λ_Y) representing the path between an observed variable Y and its η , i.e., the item [loading](#) on its [latent variable](#).
- Theta Epsilon (Θ_ϵ) representing the error variance associated with this Y item, i.e., the variance not reflecting its [latent variable](#) η .

- The holistic analysis that SEM is capable of performing is carried out via one of two distinct statistical techniques:
 1. covariance analysis
 - employed in LISREL, EQS and AMOS
 2. partial least squares
 - employed in PLS and PLS-Graph

Comparative Analysis Based on Statistics Provided by SEM

Statistics	LISREL	PLS	Regression
Analysis of overall model fit	Provided	Provided	Provided
Analysis of individual causation paths	Provided	Provided	Provided
Analysis of individual item loading paths	Provided	Provided	Not provided
Analysis of residual non-common error	Provided	Not Provided	Not provided
Type of variance examined	1. Common 2. Specific 3. Error	Common Combined specific and error	Common
Analysis of statistical power	Not available	Available through the f^2 statistic.	Available

Comparative Analysis Based on Capabilities

Capabilities	LISREL	PLS	Regression
Examines interaction effect on cause-effect paths	Supported	Supported	Supported
Examines interaction effect on item loadings	Supported	Not readily supported	Not supported
Examines interaction effect on non-common variance	Supported	Not readily supported	Not supported
Examines interaction effect on the entire model	Supported	Not readily supported	Not supported
Can cope with relatively small sample size	Problematic	Supported	Supported
Readily examines interaction effect with numerous variable levels	Problematic	Supported	Supported
Can constrain a path to a given value	Supported	Not supported	Not supported
Examines <u>nested models</u>	Supported	Supported	Supported

Comparative Analysis Based on Capabilities

Capabilities	LISREL	PLS	Regression
Establishment of causation	No	No	No
Possible over-fitting	Problematic	Less problematic	Less problematic
Testing of suspected non-linear effect	Problematic	Problematic	Mitigated by data transformation
Suspected influential outliers	Problematic	Problematic	Mitigated by data transformation
Suspected heteroscedasticity	Problematic	Problematic	Mitigated by data transformation
Suspected polynomial relation	Problematic	Problematic	Mitigated by data transformation

Heuristics for Statistical Conclusion Validity (Part 1)

Validity	Technique	Heuristic
<i>Construct Validity</i>		
Convergent Validity	CFA used in covariance-based SEM only.	GFI > .90, NFI > .90, AGFI > .80 (or >.90) and an insignificant χ^2 , to show unidimensionality . In addition, item loadings should be above .707, to show that over half the variance is captured by the latent construct [Chin, 1998b, Hair et al., 1998, Segars, 1997, Thompson et al., 1995].
Discriminant Validity	CFA used in covariance-based SEM only.	Comparing the χ^2 of the original model with an alternative model where the constructs in question are united as one construct. If the χ^2 is significantly smaller in the original model, discriminant validity has been shown [Segars, 1997].
Convergent & Discriminant Validities	PCA used in PLS can assess factor analysis but not as rigorously as a CFA in LISREL does and without examining unidimensionality	Each construct AVE should be larger than its correlation with other constructs, and each item should load more highly on its assigned construct than on the other constructs.
<i>Reliability</i>		
Internal Consistency	Cronbach's α	Cronbach's αs should be above .60 for exploratory research and above .70 for confirmatory research [Nunnally, 1967, Nunnally, 1978, Nunnally and Bernstein, 1994, Peter, 1979].
	SEM	The internal consistency coefficient should be above .70 [Hair et al., 1998, Thompson et al., 1995].
Unidimensional Reliability	Covariance-based SEM only.	Model comparisons favor unidimensionality with a significantly smaller χ^2 in the proposed measurement model in comparison with alternative measurement models [Segars, 1997].

Heuristics for Statistical Conclusion Validity (Part 2)

<i>Model Validity</i>		
AGFI	LISREL	AGFI > .80 [Segars and Grover, 1993]
Squared Multiple Correlations	LISREL, PLS	No official guidelines exist, but, clearly, the larger these values, the better
χ^2	LISREL	Insignificant and χ^2 to degrees of freedom ratio of less than 3:1 [Chin and Todd, 1995, Hair et al., 1998]
Residuals	LISREL	RMR <.05 [Hair et al., 1998]
NFI	LISREL	NFI > .90 [Hair et al., 1998]
Path Validity Coefficients	LISREL	The β and γ coefficients must be significant; standardized values should be reported for comparison purposes [Bollen, 1989, Hair et al., 1998, Jöreskog and Sörbom, 1989]
	PLS	Significant t-values [Thompson et al., 1995].
	Linear Regression	Significant t-values [Thompson et al., 1995].
<i>Nested Models</i>		
	LISREL	A nested model is rejected based on insignificant β s and γ s paths and an insignificant change in the χ^2 between the models given the change in degrees of freedom [Anderson and Gerbing, 1988] [Jöreskog and Sörbom, 1989]
	PLS	A nested model is rejected if it does not yield significant a f^2 [Chin and Todd, 1995].
	Linear Regression	A nested model in a stepwise regression is rejected if it does not yield a significant change in the F statistic (reflected directly in the change in R²) [Neter et al., 1990].

APPENDIX B

INSTRUCTIONS TO SUBJECTS AND INSTRUMENTATION

INSTRUCTIONS:

As part of an ongoing study on Internet use, we would be grateful if you could devote 10 minutes to completing this instrument.

1. Please logon to the Internet and access www.travelocity.com
2. Use the Web-site to search for a flight to Heathrow Airport (London) next month.
3. Then, please fill in the instrument below.

Please circle the appropriate category:

Gender	M , F
Age group	15-19, 20-24, 25-29, 30-34, 35-39, 40-44, 50-54, 55-59, 60-64, 65-69, above 70
What language do you speak at home (English, Italian, Hindi, Cantonese, etc.)?	
Have you ever bought products on the World Wide Web	Yes, No
How many times have you used Travelocity.com?	
Have you given your credit card number on the Web?	Yes, No

Please indicate your agreement with the next set of statements using the following rating scale:

1	2	3	4	5	6	7
Strongly Agree	Agree	Somewhat Agree	Neutral	Somewhat Disagree	Disagree	Strongly Disagree

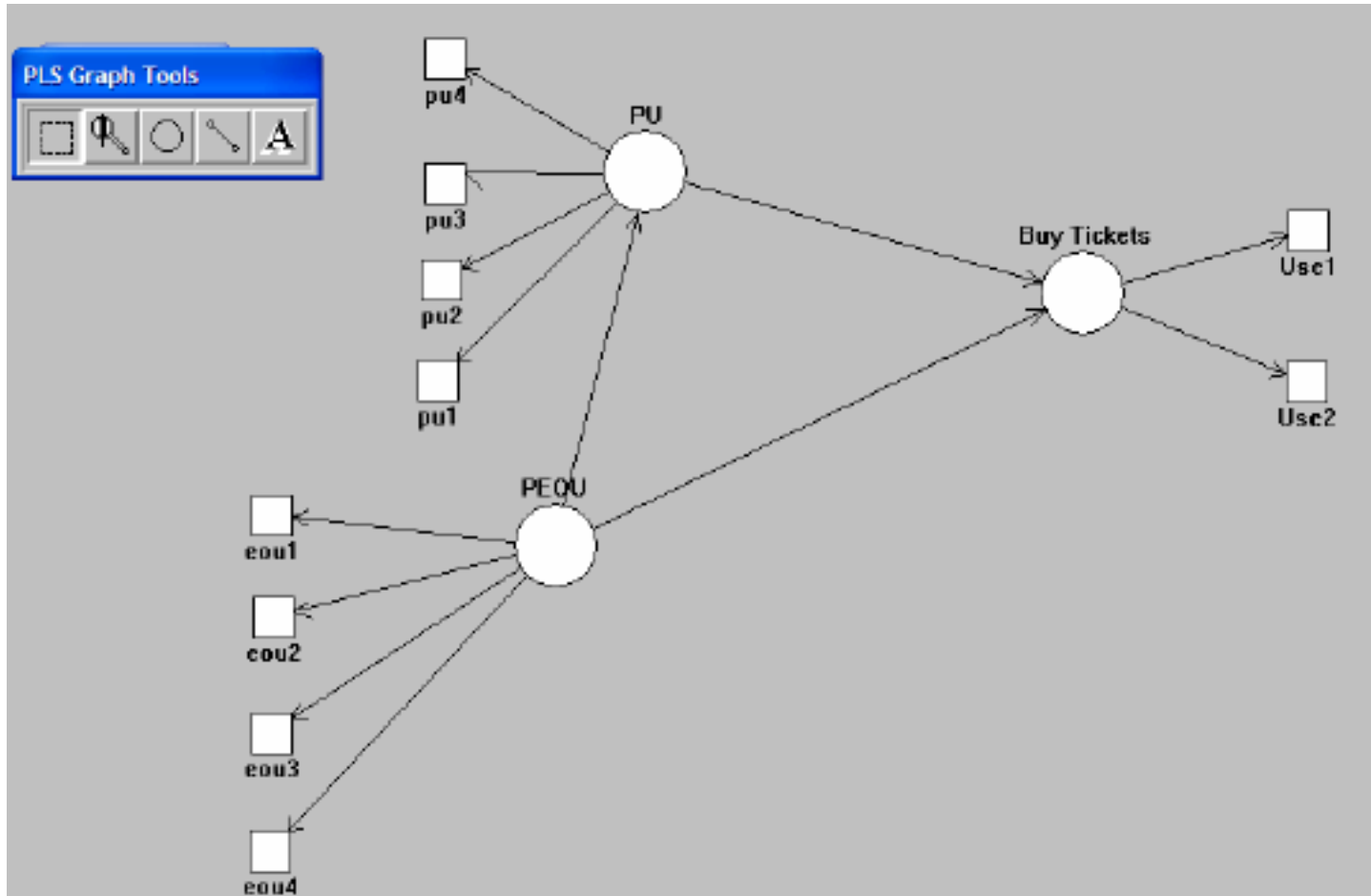
Code*	Item	Agree	Disagree
EOU1	Travelocity.com is easy to use.	1 2 3 4 5 6 7	
EOU2	It is easy to become skillful at using Travelocity.com.	1 2 3 4 5 6 7	
EOU3	Learning to operate Travelocity.com is easy .	1 2 3 4 5 6 7	
EOU4	Travelocity.com is flexible to interact with .	1 2 3 4 5 6 7	
EOU5	My interaction with Travelocity.com is clear and understandable .	1 2 3 4 5 6 7	
EOU6	It is easy to interact with Travelocity.com.	1 2 3 4 5 6 7	
PU1	Travelocity.com is useful for searching and buying flights .	1 2 3 4 5 6 7	
PU2	Travelocity.com improves my performance in flight searching and buying.	1 2 3 4 5 6 7	
PU3	Travelocity.com enables me to search and buy flights faster.	1 2 3 4 5 6 7	
PU4	Travelocity.com enhances my effectiveness in flight searching and buying.	1 2 3 4 5 6 7	
PU5	Travelocity.com makes it easier to search for and purchase flights.	1 2 3 4 5 6 7	
PU6	Travelocity.com increases my productivity in searching and purchasing flights.	1 2 3 4 5 6 7	
IUSE1	I am very likely to buy books from Travelocity.com.	1 2 3 4 5 6 7	
IUSE2	I would use my credit card to purchase from Travelocity.com.	1 2 3 4 5 6 7	
IUSE3	I would not hesitate to provide information about my habits to Travelocity.	1 2 3 4 5 6 7	

Thank You!

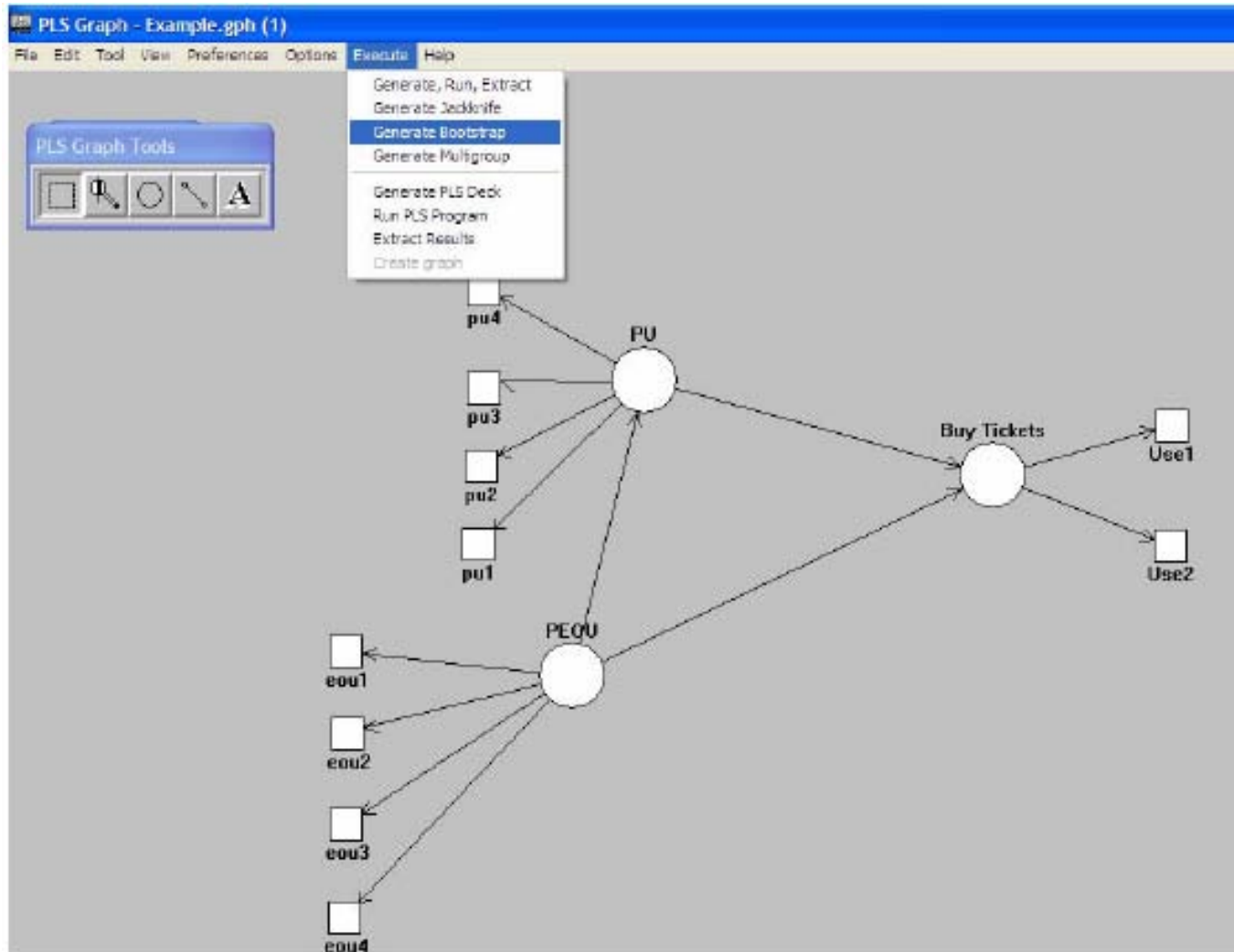
* Students did not receive the item codes****.

- Gefen, David and Straub, Detmar (2005)
"A Practical Guide To Factorial Validity Using PLS-Graph: Tutorial And Annotated Example,"
Communications of the Association for Information Systems: Vol. 16, Article 5.
Available at:
<http://aisel.aisnet.org/cais/vol16/iss1/5>

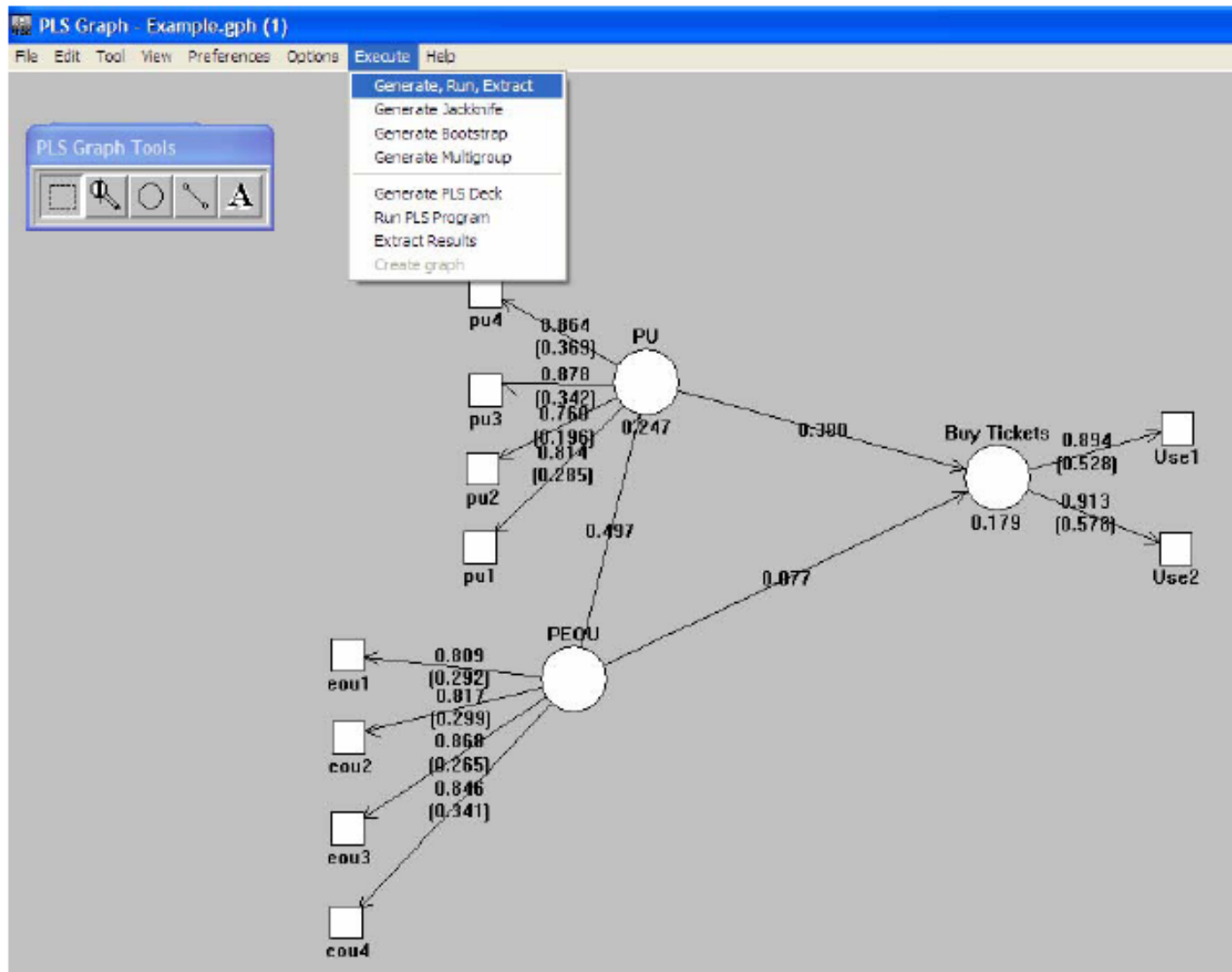
PLS-Graph Model



Extracting PLS-Graph Model



Displaying the PLS-Graph Model



PCA with a Varimax Rotation of the Same Data

	Component		
	1	2	3
eou3	.894	.092	.072
eou2	.784	.178	.115
eou1	.782	.167	.114
eou4	.771	.310	.047
pu2	.097	.856	-.034
pu1	.159	.810	.164
pu3	.261	.772	.260
pu4	.337	.700	.294
Use1	.030	.186	.883
Use2	.186	.144	.870

Extraction Method: Principal Component Analysis.
 Rotation Method: Varimax with Kaiser Normalization.
 Rotation converged in 5 iterations.

Correlations in the 1st file as compared with the Square Root of the AVE

Correlations of latent variables

	Buy Tick	PU	PEOU
Buy Tick	1.000		
PU	0.418	1.000	
PEOU	0.266	0.497	1.000

	AVE	SQRT of AVE
Buy Ticket	0.817	0.903881
PU	0.69	0.830662
PEOU	0.698	0.835464

Summary

- Confirmatory Factor Analysis (CFA) & Structured Equation Modeling (SEM)
- Covariance based SEM
 - LISREL
- Partial-least-squares (PLS) based SEM
 - PLS

References

- Joseph F. Hair, William C. Black, Barry J. Babin, Rolph E. Anderson (2009), *Multivariate Data Analysis*, 7th Edition, Prentice Hall
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