



My Research

KMIS Doctoral Consortium 2006



Bock, Gee-Woo



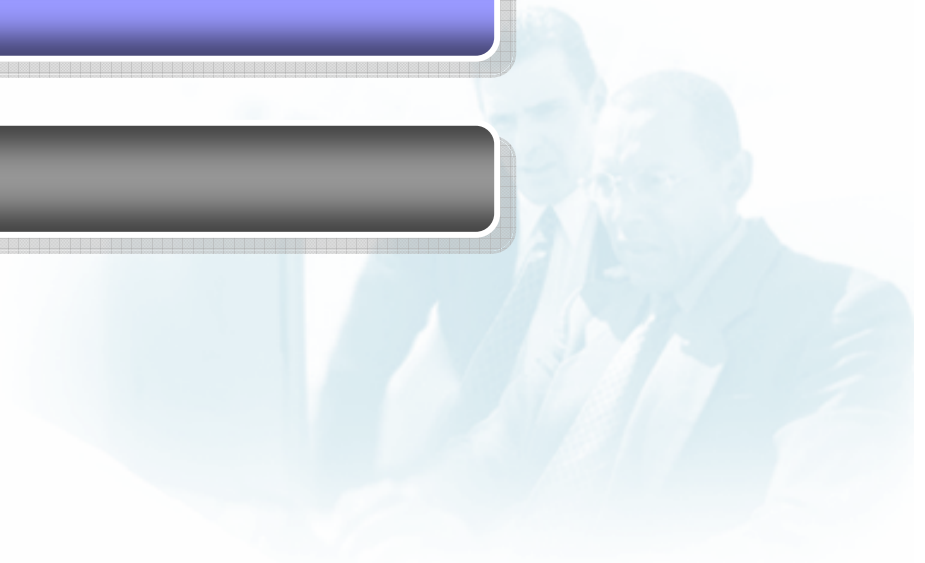
Agenda

1. Research Interests

2. How to execute

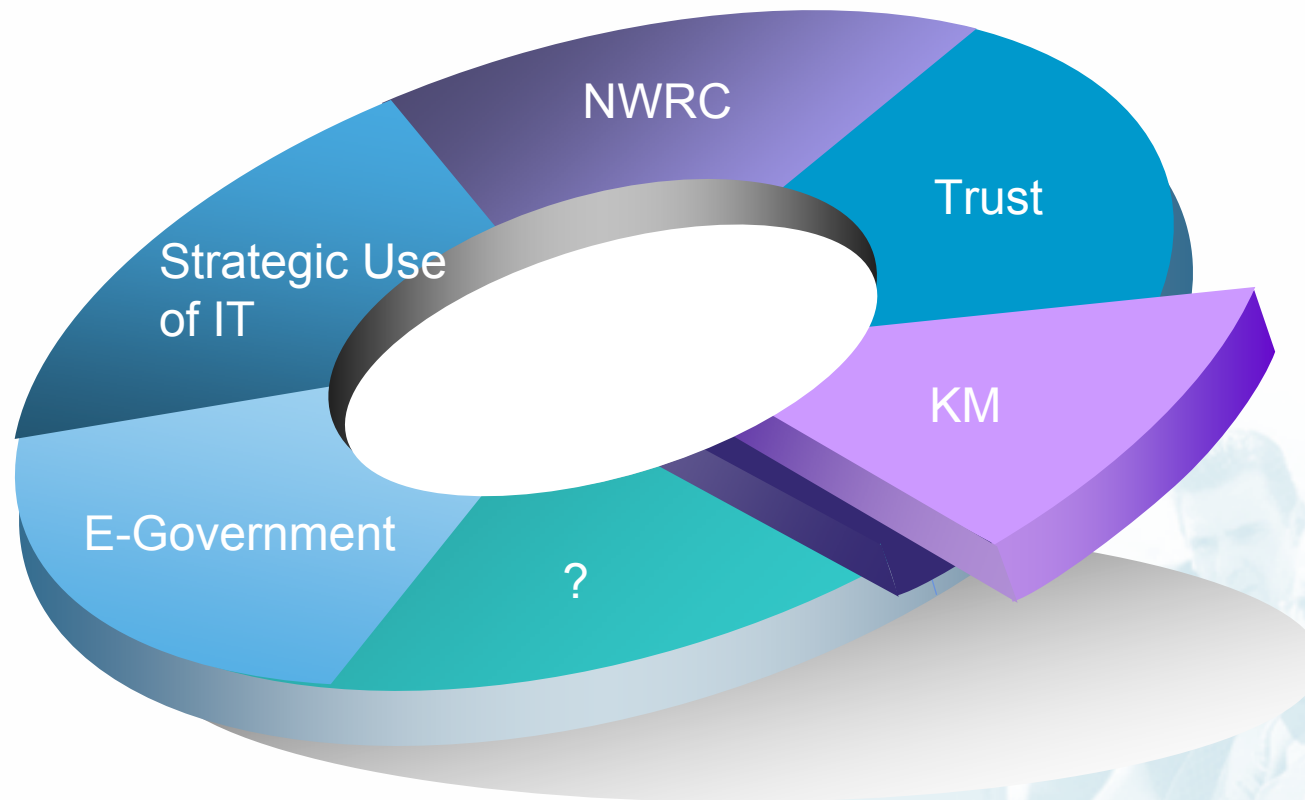
3. How to publish

4. Q&A



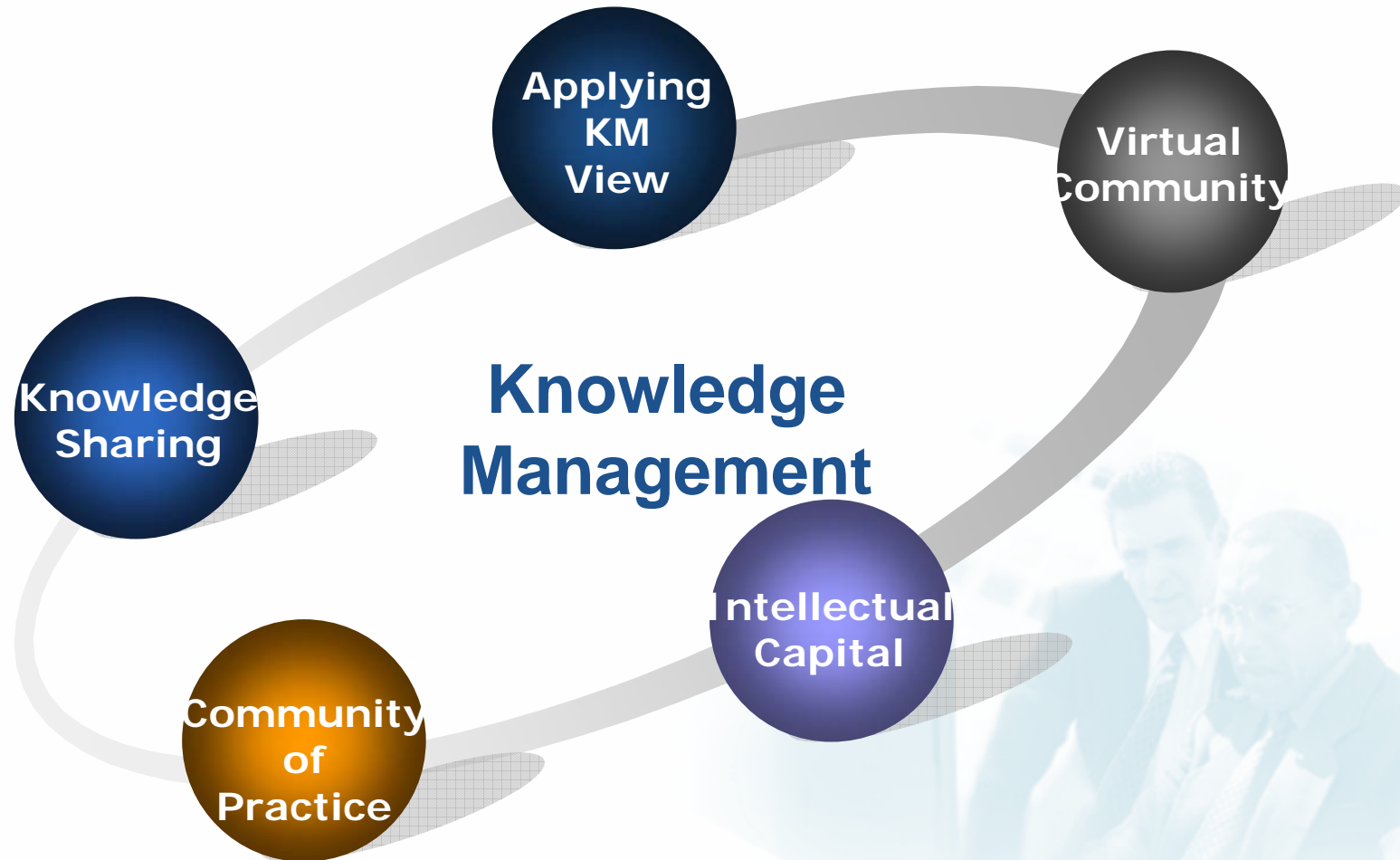


Research Interests





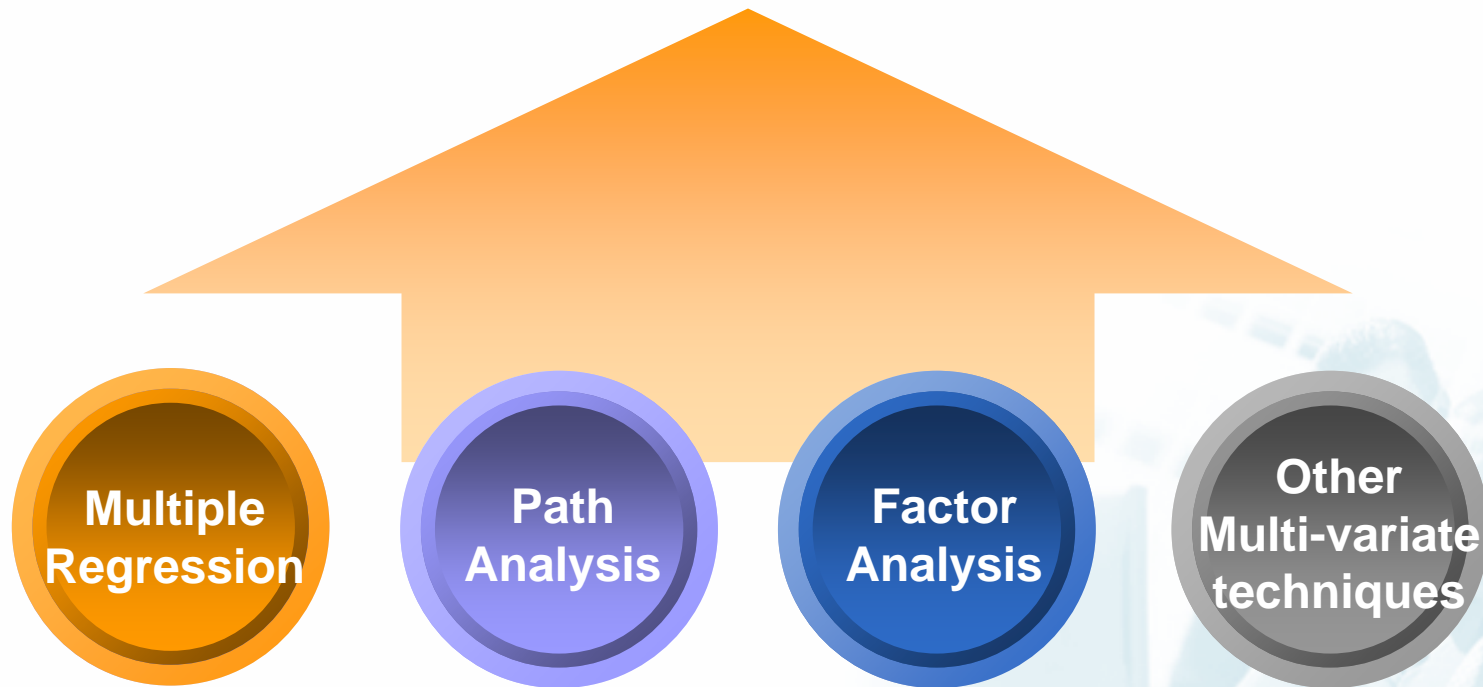
Research Issues in Knowledge Management





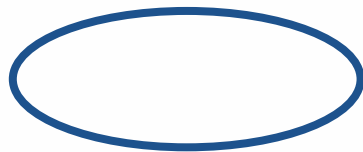
How to execute

Structural Equation Modeling

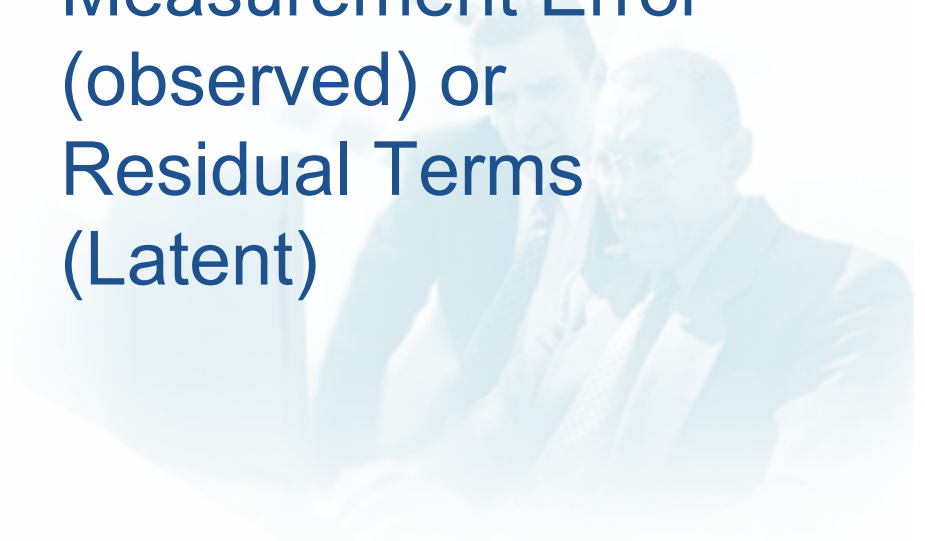




Symbols Used in SEM

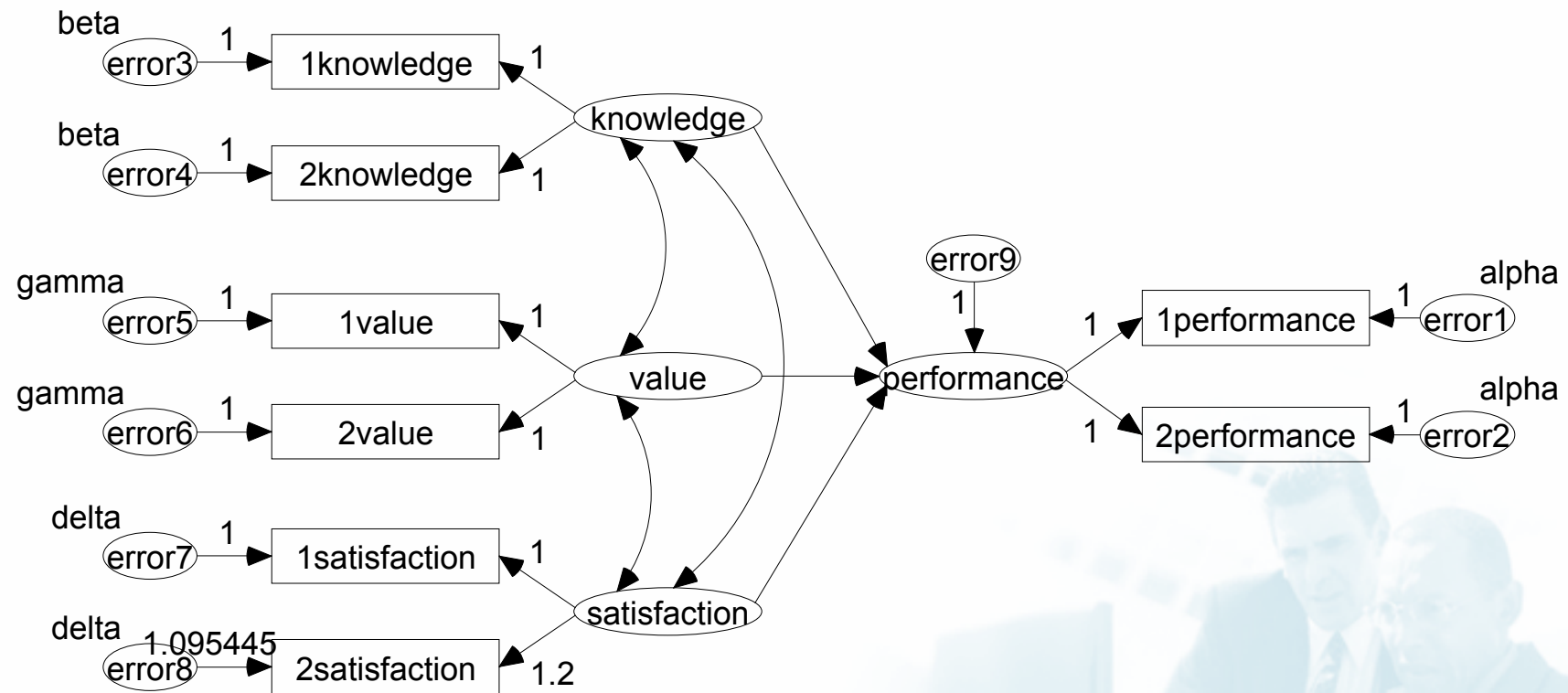


- Observed Variables
- Latent (unobserved) Variables
- Measurement Error (observed) or Residual Terms (Latent)





MR, PA & FA within SEM



Example 5: Model B
Parallel tests regression
Job performance of farm managers
Warren, White and Fuller (1974)
Model Specification



SEM is...

- Fantastic for theory-testing
- Able to overcome 'problems' with conventional analytical techniques, such as:
 - MR does not allow exploration of relationships between dependent variables
 - MR, path analysis use single indicator variables as proxies for complex constructs
 - When using composite variables, it is assumed that all indicators are measured with equal error and all contribute equally



Continued...

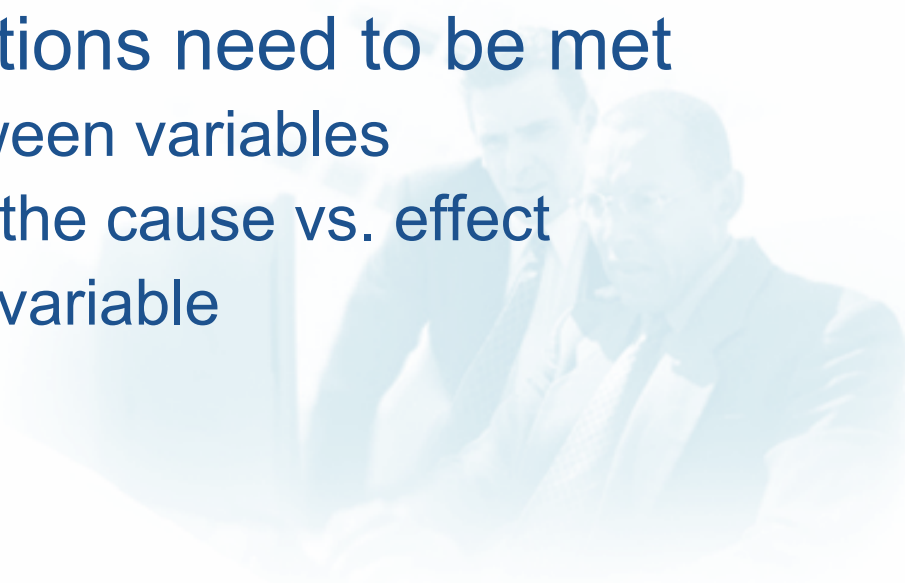
- Analysis of covariance structure
- Latent variable analysis
- Causal modeling
- LISREL
 - PLS, AMOS





Cautionary note!

- SEM does not prove causality between variables!
 - SEM, like other relational techniques, is based on dependence relationships (change in one variable results in change in another). However, causation is dependent on theory and study design, not analytical technique.
- For causation, four conditions need to be met
 - Sufficient association between variables
 - Temporal antecedence of the cause vs. effect
 - Lack of alternative causal variable
 - Theoretical bases



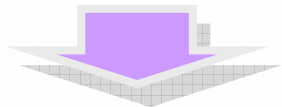


How does SEM work?

- In a nutshell, if $S = \Sigma$, then the model fits the data!
 - S = Empirical/observed/sample variance/covariance matrix
 - Σ = Model implied variance/covariance matrix

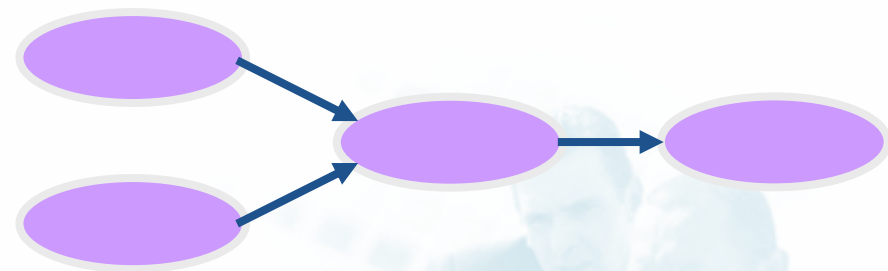
S – from Data

x1	x2	y
1	2	3
3	4	4
4	5	2
3	1	5



Var(x1)		
Cov(x1,x2)	Var(x2)	
Cov(x1,y)	Cov(x2,y)	Var(y)

Σ – from Model



0		
a+bx		
0	by+c	0

Are these matrices
'statistically' equal?





SEM Analysis Process

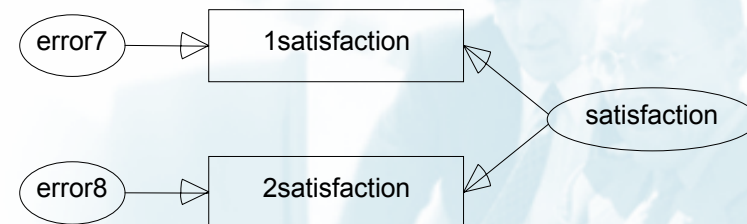
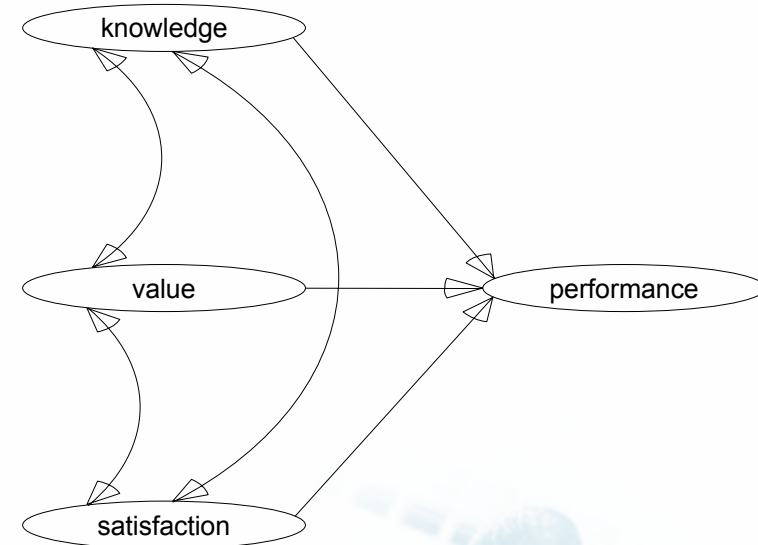
- Stage 1: Specify model
- Stage 2: Determine whether the model is identified
- Stage 3: Select measures of variables in the model and collect data
- Stage 4: Analyze model
- Stage 5: Evaluate model-data fit
- Stage 6: If necessary, re-specify model and re-evaluate revised model





Stage 1: Specify model

- Theory based structural model
- How will constructs be measured? Measurement models
- Full specification: Measurement + Structural Models
 - Single-step approach
 - Two-step approach
 - Theory is not well established
 - Measurement instrument is untested

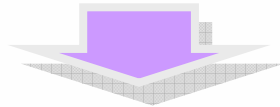




Stage 2: Determine whether model is identified

S – from Data

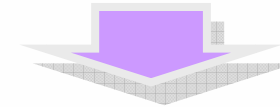
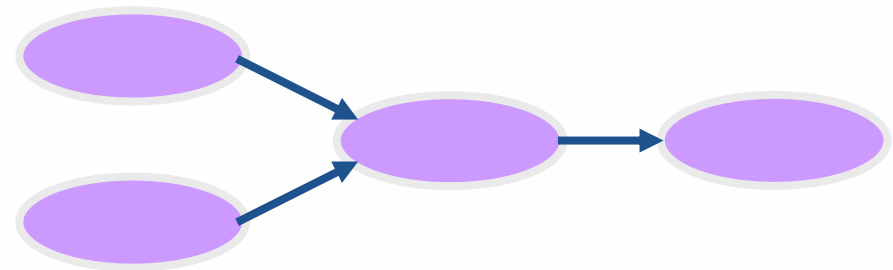
x1	x2	y
1	2	3
3	4	4
4	5	2
3	1	5



Var(x1)		
Cov(x1,x2)	Var(x2)	
Cov(x1,y)	Cov(x2,y)	Var(y)

Total # of
observations:
 $v(v+1)/2$

Σ – from Model



0		
a+bx		
0	by+c	0

Total # of
parameters:
t

- If $v(v+1)/2 < t$, too many variables and not enough equations. SEM cannot proceed: Model is under-identified.
- If $v(v+1)/2 = t$, then unique values for free parameters. SEM is unnecessary: Model is just-identified.
- If $v(v+1)/2 > t$, multiple values for each free parameters. SEM ideally suited: Model is over-identified.
- N.B. degree of freedom (df) = $[v(v+1)/2] - t$



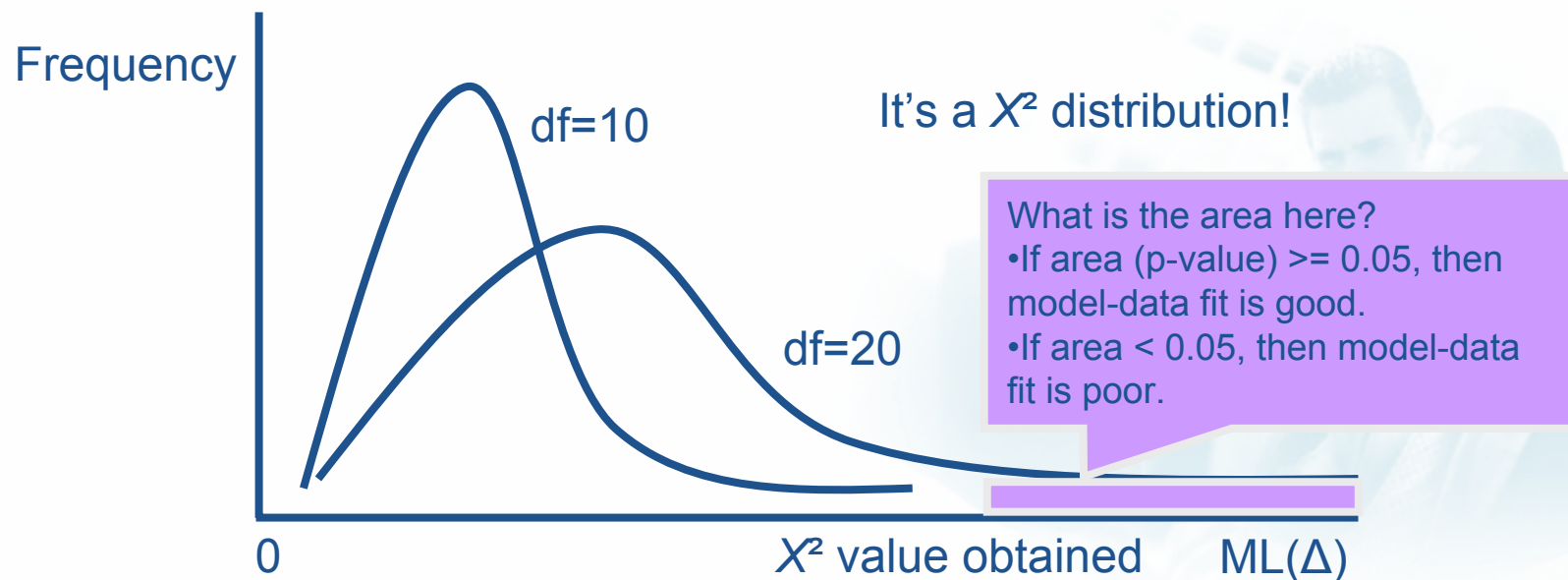
Stage 4: Analyze model

- Parameter calculations
 - If more than 1 solution for free parameters, then use systematic procedure to select appropriate value.
 - Method available
 - Simple average
 - Maximum likelihood
 - Generalized least squares etc.
 - E.g. $\text{Var}(x) = \{1, 5, 6\}$
 - Average = 4
 - ML = 4.1?
 - GLS = 3.8?
- Calculate difference between observed and model-implied matrices
 - $S - \Sigma = \Delta$ ('Discrepancy' function)
- Iteratively minimize Δ
 - If $\Delta = 0$, then model-data fit is good. Go to Stage 5.
 - If $\Delta > 0$, then model-data fit is poor. Iteratively update values in model-implied matrix, until Δ is minimized.



Stage 5: Evaluate model-data fit

- Δ , in matrix form, is not very 'user-friendly'
- Convert Δ to 'usable' statistic using same parameter estimation technique, e.g.
 - Simple average $\Delta = \text{Av}(a_{ij})$
 - Maximum likelihood $\Delta = \log|\Sigma| + \text{tr}(S\Sigma^{-1}) - \log|S| - (p+q)$
- Draw frequency distribution





Continued...

- Problems with χ^2 -based measures:
 - Sensitive to large sample sizes (i.e., always say poor fit if $n > 200$)
 - No upper bound – so not easy to interpret in a standard way
- To overcome these problems, many other fit statistics have been proposed.
 - Absolute Fit Indices
 - Incremental Fit Indices
 - Model Parsimony





Stage 6: If necessary, Re-specify model & Re-evaluate revised model

- When model-data fit is poor, re-specify model and re-evaluate model (i.e. return to step 1)
- Look for ideas for model improvement by reviewing modification indices and residuals
- Caution: This step is controversial – some regard it as post-hoc revisionism!





Additional Points to Note

- Data type:
 - Best if all measures are interval scale
 - If ordinal scale (Likert), SEM results are conservative (i.e., some effects will not be detected)
- Model complexity
 - Keep it simple!
 - Sample size restrictions – 10 cases/parameter
- SEM is excellent for testing mediating effects
- SEM is not so good for moderating effects
 - How to standardize (normalize)
 - Multiply the 2 constructs together using excel.
 - Then open the file in SPSS.
 - Analyze - descriptive statistics - descriptives.
 - Select the multiplied items and save standardized values as variables.
 - The standardized values will appear.
 - Then use these values as the moderating variable in SEM.



Conclusions

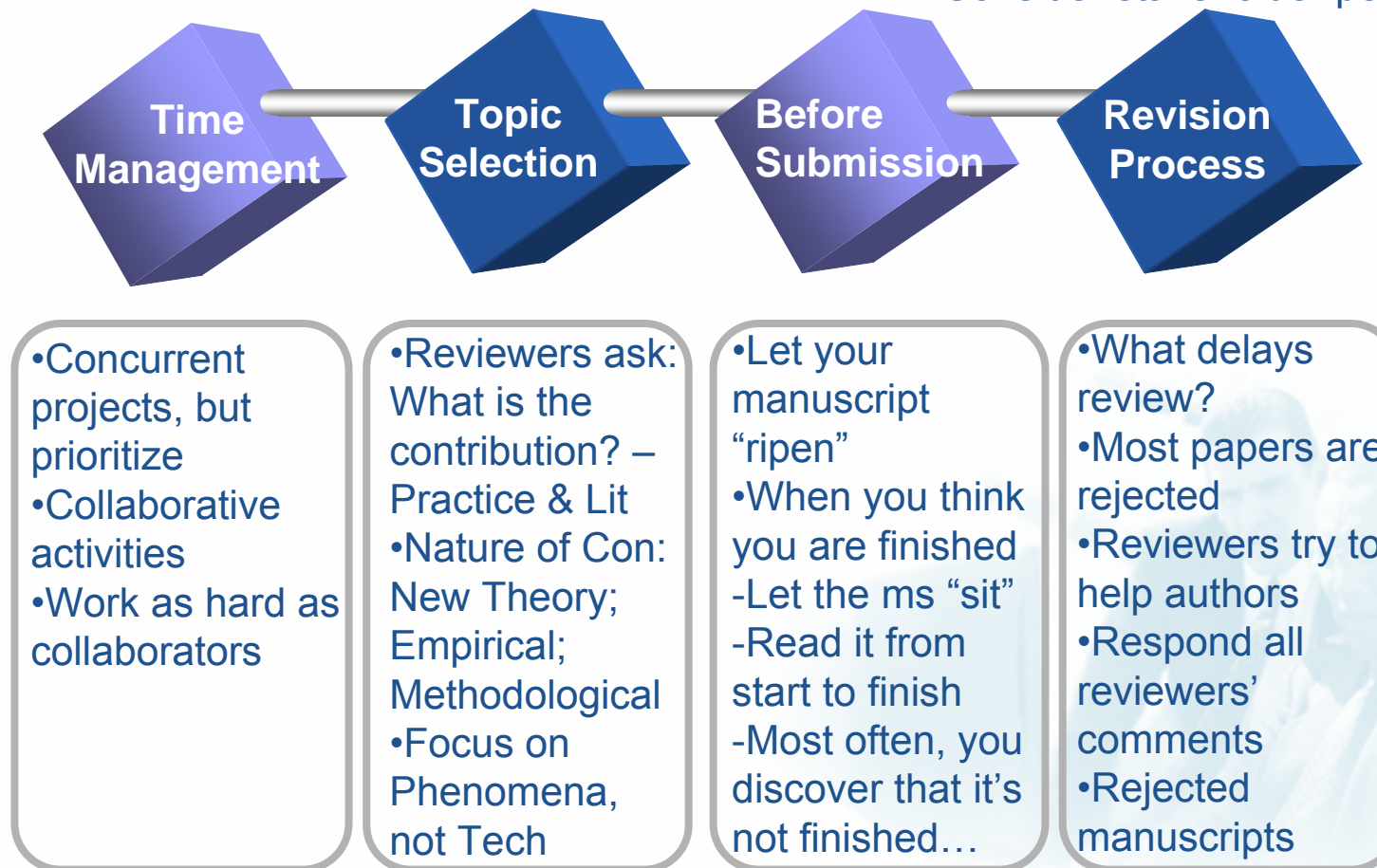
- SEM is a very 'powerful' method for testing models (but requires a lot of care and attention detail!)
 - It depends on 'good' theory!
 - It depends on 'good' data!
- There may be other more appropriate methods (such as MR) for testing models.





How to publish in top journals

- Choose a right journal
- Help reviewers
- Consider stakeholder perspective





Crafting Your Manuscript

Introduction

Describe
Convince
Provide
a “roadmap”

Theory

Articulate
Bridge
Provide
a conceptual
framework
Keep consistency
Present prior
literature...
Precede
hypotheses..

Methods

Need to know
what you have
done
Wish to apply
yours
Justify
Describe
Hide

Discussion

Describe
analyses &
findings
Point out
Interpret
Use conceptual
model;
anecdotes;
figures, tables;
Don't overstate

Conclusion:

What you should not do...summarize the article, introduce anything for the first time, discuss weaknesses
What you should do...focus on key messages, market the contributions, maintain a positive tone