

NAME (PRINT): _____
Last/Surname First /Given Name

STUDENT #: _____ SIGNATURE: _____

UNIVERSITY OF TORONTO MISSISSAUGA
APRIL 2011 FINAL EXAMINATION
STA431H5S
Structural Equation Models
Jerry Brunner
Duration - 3 hours
Aids: Statistical Calculators; Reference Sheet (Supplied)

The University of Toronto Mississauga and you, as a student, share a commitment to academic integrity. You are reminded that you may be charged with an academic offence for possessing any unauthorized aids during the writing of an exam, including but not limited to any electronic devices with storage, such as cell phones, pagers, personal digital assistants (PDAs), iPods, and MP3 players. Unauthorized calculators and notes are also not permitted. Do not have any of these items in your possession in the area of your desk. Please turn the electronics off and put all unauthorized aids with your belongings at the front of the room before the examination begins. If any of these items are kept with you during the writing of your exam, you may be charged with an academic offence. A typical penalty may cause you to fail the course.

*Please note, you **CANNOT** petition to **re-write** an examination once the exam has begun.*

Qn. #	Value	Score
1	10	
2	27	
3	23	
4	12	
5	10	
6	18	
Total = 100 Points		

10 points

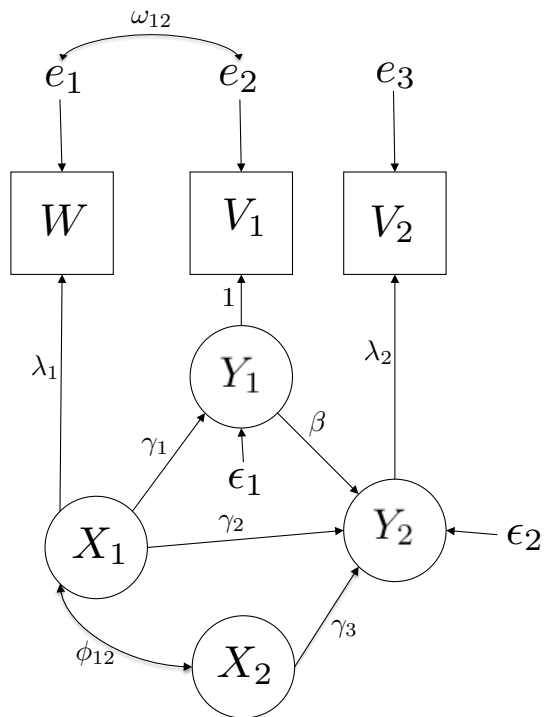
1. Let F be a latent variable. Consider two equivalent measurements with uncorrelated measurement error:

$$\begin{aligned}D_1 &= \nu + F + e_1 \\D_2 &= \nu + F + e_2,\end{aligned}$$

where $E(F) = \mu$, $Var(F) = \phi$, $E(e_1) = E(e_2) = 0$, $Var(e_1) = Var(e_2) = \omega$, and F , e_1 and e_2 are all independent. The observable variables in this model are D_1 and D_2 . The *reliability* of an observable measurement is defined as the squared correlation between the true (latent) measurement and the observable measurement. Show that the reliability of D_1 equals the correlation between D_1 and D_2 . Hint: both quantities equal a familiar formula.

27 points

2. Consider the following path diagram.



(a) Write down the model equations in scalar form.

(b) Refer to the general two-stage structural equation model on the formula sheet. Write down the following, using symbols from the path diagram. This part is intended to help you use matrices with the correct dimensions.

i. \mathbf{X}

ii. \mathbf{Y}

iii. \mathbf{F}

iv. \mathbf{D}

(c) Give the matrix β , using symbols from the path diagram. *Make sure it has the correct dimensions.*

- (d) Give the matrix $\mathbf{\Gamma}$, using symbols from the path diagram. Make sure it has the correct dimensions.
- (e) Give the matrix $\mathbf{\Lambda}$, using symbols from the path diagram. Make sure it has the correct dimensions.
- (f) Give the matrix $\mathbf{\Phi}_{11}$, using symbols from the path diagram. Make sure it has the correct dimensions. Not all symbols in $\mathbf{\Phi}_{11}$ appear on the path diagram.
- (g) Give the matrix $\mathbf{\Psi}$. Make sure it has the correct dimensions. The symbols in $\mathbf{\Psi}$ do not appear on the path diagram. Use standard notation.
- (h) Give the matrix $\mathbf{\Omega}$, using symbols from the path diagram. Make sure it has the correct dimensions. Not all symbols in $\mathbf{\Omega}$ appear on the path diagram.

23 points

3. For the measurement part of the general two-stage structural equation model (see formula sheet), suppose Φ is the identity and Ω is diagonal. In addition, suppose that the observable variables have been standardized, and we write \mathbf{Z} in place of \mathbf{D} . This yields the classical exploratory factor analysis model. In your calculations, assume \mathbf{F} is $p \times 1$ and \mathbf{Z} is $k \times 1$.

(a) Show that factor loadings are actually correlations by calculating $\text{Corr}(Z_i, F_j)$. This is a *scalar* (not matrix) calculation. Show your work. *Circle your final answer.*

(b) Show that $\omega_{i,i} = \text{Var}(e_i)$ is a function of the $\lambda_{i,j}$ quantities. This is a *scalar* (not matrix) calculation. Show your work. *Circle your final answer.*

(c) What are the *unknown* parameters in this classical factor analysis model? The answer is a list of one or more matrices.

- (d) Prove that the parameters of the the classical exploratory factor analysis model (the model you are working with in this question) are not identifiable. This is the case even if there are enough observable variables so that the counting rule is satisfied, so don't try to use the counting rule. Show your work.

12 points

4. let

$$\begin{aligned}D_1 &= \lambda_1 F + e_1 \\D_2 &= \lambda_2 F + e_2 \\D_3 &= \lambda_3 F + e_3,\end{aligned}$$

where all expected values are zero, $V(e_i) = \omega_i$ for $i = 1, 2, 3$, $V(F) = \phi$, the error terms are independent of one another and of the factor, $\lambda_1 > 0$, $\lambda_2 \neq 0$ and $\lambda_3 \neq 0$. The parameters of the model are not identifiable (you don't have to show this), but you can work with it anyway; *do not re-parameterize!*

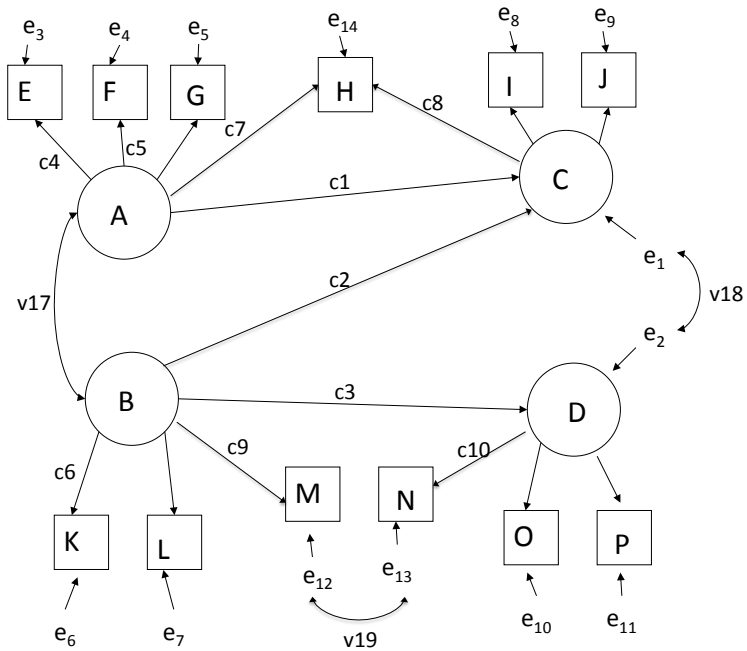
(a) Give the covariance matrix Σ . This is a scalar calculation.

(b) Show that the ratio of factor loadings λ_1/λ_2 is identifiable.

- (c) Show that the parameter ω_1 is identifiable. You have more room on this page than you need.

10 points

5. The parameters of the model represented by the following path diagram are identifiable. Briefly explain why. Make specific reference to the identifiability rules on the reference sheet.



18 points

6. The *Poverty Data* contain information from a sample of 97 countries. The variables include Live birth rate per 1,000 of population, Death rate per 1,000 of population, Infant deaths per 1,000 of population under 1 year old, Life expectancy at birth for males, Life expectancy at birth for females, and Gross National Product per capita in U.S. dollars. Please answer these questions based on the SAS program and list file starting on Page 12.

Note: The printout shows two runs of `proc calis`. Each one fits a different model.

- (a) What is the parameter θ for the factor analysis model? That's the first run of `proc calis`. Give your answer in the form of a list of names from the SAS job.

- (b) Does the model fit the data adequately? Answer Yes or No, and back up your answer with two numbers from the printout: The value of a test statistic, and a p -value.

- (c) What is the maximum likelihood estimate of the correlation between factors? The answer is a single number from the printout.

- (d) Clearly, the factor loadings are not identifiable for points in the parameter space where $\phi_{12} = 0$. In your view, is this likely to be a problem in the present case? Answer "Yes, a problem," or "No problem" and briefly explain. Does the printout support your position?

- (e) Now consider the second run of `proc calis` – the latent regression model. Does this model fit the data adequately? Answer Yes or No. Is the fit better than the factor analysis model, worse than the factor analysis model, or is it the same? Write one of these alternatives and *circle it*.

- (f) According to the regression model, wealth causes health if there is any connection between the two. How do you know *from the printout* that more money translates into better health (and not worse health)? Cite a particular number from the printout.

That's the end of the questions. The rest of this exam script is computer printout.

Total Marks = 100 points

continued on page 12

```

/***** finalpoverty2011B.sas *****/
options linesize=79 noovp formdlim='-';
title 'UN Poverty Data: STA431 S 2011 Final Exam';

data misery;
  infile 'poverty.data';
  input birthrate deathrate infmort lifexM lifexF gnp group country $;
  gnp1000 = gnp/1000; /* In thousands of dollars */
  lifex = (lifexM+lifexF)/2;

proc calis cov;
  title2 'Factor analysis model standardizing the factors';
  var lifex infmort gnp1000 birthrate;
  lineqs
    lifex      = lambda1 Fhealth + e1,
    infmort    = lambda2 Fhealth + e2,
    gnp1000    = lambda3 Fwealth + e3,
    birthrate  = lambda4 Fwealth + e4;
  std
    Fhealth = 1, Fwealth=1,
    e1-e4 = 4 * omega: ;
  cov Fhealth Fwealth = phi12;
  bounds 0.0 < omega1-omega4;

proc calis cov;
  title2 'Latent regression model setting loadings to one';
  var lifex infmort gnp1000 birthrate;
  lineqs
    Fhealth    = gamma Fwealth + epsilon,
    lifex      =          Fhealth + e1,
    infmort    = lambda2 Fhealth + e2,
    gnp1000    =          Fwealth + e3,
    birthrate  = lambda4 Fwealth + e4;
  std
    Fwealth=phi, epsilon = psi,
    e1-e4 = 4 * omega: ;
  bounds 0.0 < phi psi omega1-omega4;

```


UN Poverty Data: STA431 S 2011 Final Exam 6
 Factor analysis model standardizing the factors

The CALIS Procedure
 Covariance Structure Analysis: Maximum Likelihood Estimation

Fit Function	0.0121
Goodness of Fit Index (GFI)	0.9940
GFI Adjusted for Degrees of Freedom (AGFI)	0.9404
Root Mean Square Residual (RMR)	2.5734
Standardized Root Mean Square Residual (SRMR)	0.0067
Parsimonious GFI (Mulaik, 1989)	0.1657
Chi-Square	1.0862
Chi-Square DF	1
Pr > Chi-Square	0.2973
Independence Model Chi-Square	399.83
Independence Model Chi-Square DF	6

UN Poverty Data: STA431 S 2011 Final Exam 7
 Factor analysis model standardizing the factors

The CALIS Procedure
 Covariance Structure Analysis: Maximum Likelihood Estimation

Manifest Variable Equations with Estimates

lifex	=	10.3124*	Fhealth	+	1.0000	e1
Std Err		0.7841	lambda1			
t Value		13.1522				
infmort	=	-44.2690*	Fhealth	+	1.0000	e2
Std Err		3.6165	lambda2			
t Value		-12.2408				
gnp1000	=	5.4842*	Fwealth	+	1.0000	e3
Std Err		0.7664	lambda3			
t Value		7.1560				
birthrate	=	-12.7179*	Fwealth	+	1.0000	e4
Std Err		1.1530	lambda4			
t Value		-11.0302				

Variances of Exogenous Variables

Variable	Parameter	Estimate	Standard Error	t Value
Fhealth		1.00000		
Fwealth		1.00000		
e1	omega1	1.44765	1.84498	0.78
e2	omega2	184.16184	43.51831	4.23
e3	omega3	35.43109	5.57604	6.35
e4	omega4	25.92052	10.36055	2.50

Covariances Among Exogenous Variables

Var1	Var2	Parameter	Estimate	Standard Error	t Value
Fhealth	Fwealth	phi12	0.96065	0.03051	31.49

 UN Poverty Data: STA431 S 2011 Final Exam 9
 Latent regression model setting loadings to one

The 5 Endogenous Variables

Manifest	lifex	infmort	gnp1000	birthrate
Latent	Fhealth			

The 6 Exogenous Variables

Manifest					
Latent	Fwealth				
Error	epsilon	e1	e2	e3	e4

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 Latent regression model setting loadings to one

Optimization Start

Active Constraints 0 Objective Function 0.0176241672
 Max Abs Gradient Element 0.0096430636 Radius 1

Iter	Rest arts	Func Calls	Act Con	Objective Function	Obj Fun Change	Max Abs Gradient Element	Lambda	Actual Over Pred Change
1	0	2	0	0.01216	0.00547	0.00296	0	1.097
2	0	3	0	0.01207	0.000086	0.000199	0	1.019
3	0	4	0	0.01207	1.064E-6	0.000012	0	1.018
4	0	5	0	0.01207	1.596E-8	3.008E-6	0	1.017

Optimization Results

Iterations	4	Function Calls	6
Jacobian Calls	5	Active Constraints	0
Objective Function	0.0120694315	Max Abs Gradient Element	3.0080773E-6
Lambda	0	Actual Over Pred Change	1.0166650766
Radius	0.0010397183		

ABSGCONV convergence criterion satisfied.

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 Latent regression model setting loadings to one

The CALIS Procedure
 Covariance Structure Analysis: Maximum Likelihood Estimation

Fit Function	0.0121
Goodness of Fit Index (GFI)	0.9940
GFI Adjusted for Degrees of Freedom (AGFI)	0.9404
Root Mean Square Residual (RMR)	2.5734
Standardized Root Mean Square Residual (SRMR)	0.0067
Parsimonious GFI (Mulaik, 1989)	0.1657
Chi-Square	1.0862
Chi-Square DF	1
Pr > Chi-Square	0.2973
Independence Model Chi-Square	399.83
Independence Model Chi-Square DF	6

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 Latent regression model setting loadings to one

The CALIS Procedure
 Covariance Structure Analysis: Maximum Likelihood Estimation

Manifest Variable Equations with Estimates

lifex	=	1.0000	Fhealth	+	1.0000	e1
infmort	=	-4.2928*	Fhealth	+	1.0000	e2
Std Err		0.1650	lambda2			
t Value		-26.0109				
gnp1000	=	1.0000	Fwealth	+	1.0000	e3
birthrate	=	-2.3190*	Fwealth	+	1.0000	e4
Std Err		0.2961	lambda4			
t Value		-7.8315				

UN Poverty Data: STA431 S 2011 Final Exam
 Latent regression model setting loadings to one

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The CALIS Procedure
 Covariance Structure Analysis: Maximum Likelihood Estimation

Latent Variable Equations with Estimates

Fhealth = 1.8064*Fwealth + 1.0000 epsilon
 Std Err 0.2317 gamma
 t Value 7.7958

Variances of Exogenous Variables

Variable	Parameter	Estimate	Standard Error	t Value
Fwealth	phi	30.07656	8.40592	3.58
epsilon	psi	8.20441	6.16622	1.33
e1	omega1	1.44764	1.84498	0.78
e2	omega2	184.16198	43.51832	4.23
e3	omega3	35.43109	5.57604	6.35
e4	omega4	25.92052	10.36055	2.50