

## STA 431s11 Assignment 5

1. Here is a multivariate regression model with no intercept and no measurement error. Independently for  $i = 1, \dots, n$ ,

$$\mathbf{Y}_i = \boldsymbol{\beta}\mathbf{X}_i + \boldsymbol{\epsilon}_i$$

where

$\mathbf{Y}_i$  is an  $q \times 1$  random vector of observable dependent variables, so the regression can be multivariate; there are  $q$  dependent variables.

$\mathbf{X}_i$  is a  $p \times 1$  multivariate normal observable random vector; there are  $p$  independent variables.  $\mathbf{X}_i$  has expected value zero and variance-covariance matrix  $\boldsymbol{\Phi}$ , a  $p \times p$  symmetric and positive definite matrix of unknown constants.

$\boldsymbol{\beta}$  is a  $q \times p$  matrix of unknown constants. These are the regression coefficients, with one row for each dependent variable and one column for each independent variable.

$\boldsymbol{\epsilon}_i$  is the error term of the latent regression. It is an  $q \times 1$  multivariate normal random vector with expected value zero and variance-covariance matrix  $\boldsymbol{\Psi}$ , a  $q \times q$  symmetric and positive definite matrix of unknown constants.  $\boldsymbol{\epsilon}_i$  is independent of  $\mathbf{X}_i$ .

- (a) Calculate the variance-covariance matrix of the observable variables. It's a partitioned matrix. Show your work.
  - (b) Write down the moment structure equations. These are matrix equations.
  - (c) Are the parameters of this model identifiable? Answer Yes or No and prove your answer.
2. This question shows what can happen when errors of measurement have a non-zero covariance. Independently for  $i = 1, \dots, n$ , let

$$\begin{aligned} Y_i &= \beta X_i + \epsilon_i \\ W_{i,1} &= X_i + e_{i,1} \\ W_{i,2} &= X_i + e_{i,2}, \end{aligned}$$

where

- $X_i$  is a normally distributed *latent* variable with mean zero and variance  $\phi > 0$
- $\epsilon_i$  is normally distributed with mean zero and variance  $\psi > 0$
- $e_{i,1}$  is normally distributed with mean zero and variance  $\omega_{1,1} > 0$
- $e_{i,2}$  is normally distributed with mean zero and variance  $\omega_{2,2} > 0$
- $Cov(e_{i,1}, e_{i,2}) = \omega_{1,2}$ . This is the unusual feature (unusual in statistical models – maybe not so unusual in reality).
- $X_i$  and  $\epsilon_i$  are independent of one another.
- $X_i$  is independent of  $e_{i,1}$  and  $e_{i,2}$
- $\epsilon_i$  is independent of  $e_{i,1}$  and  $e_{i,2}$ .

- (a) What is the parameter vector  $\boldsymbol{\theta}$  for this model?
- (b) Does this problem pass the test of the Counting Rule? Answer Yes or No.

- (c) Calculate the variance-covariance matrix of the observable variables. Remember that  $Cov(e_{i,1}, e_{i,2}) \neq 0$ . Show your work.
- (d) There are 6 covariance structure equations in 6 unknowns. Try to solve them. If you can do it, you have proved that the parameter is identifiable, and you are done.
- (e) If you cannot solve the covariance structure equations, try to prove that the parameter vector is *not* identifiable. To do this, you need a simple numerical example: two different  $\theta$  vectors that produce the same  $\Sigma$ . To make it easier on yourself, let  $\beta = 0$  in both vectors. Be sure to give the covariance matrix (a  $3 \times 3$  matrix of numbers) that is produced by both sets of parameter values. In your example, make sure  $|\Omega| > 0$  (a point that is easy to overlook).
3. Consider the following simple regression through the origin with measurement error in both the independent and dependent variables. Independently for  $i = 1, \dots, n$ ,

$$\begin{aligned} Y_i &= \beta X_i + \epsilon_i \\ W_{i,1} &= X_i + e_{i,1} \\ V_{i,1} &= Y_i + e_{i,2} \\ W_{i,2} &= X_i + e_{i,3} \\ V_{i,2} &= Y_i + e_{i,4} \end{aligned}$$

where

- $X_i$  and  $Y_i$  are latent variables
- $X_i \sim N(0, \phi)$
- $\epsilon_i \sim N(0, \psi)$
- $\mathbf{e}_i = (e_{i,1}, e_{i,2}, e_{i,3}, e_{i,4})'$
- $X_i$ ,  $\epsilon_i$  and  $\mathbf{e}_i$  are independent of one another
- $\mathbf{e}_i$  is multivariate normal with mean zero and covariance matrix

$$\Omega = \begin{bmatrix} \omega_{1,1} & \omega_{1,2} & 0 & 0 \\ \omega_{1,2} & \omega_{2,2} & 0 & 0 \\ 0 & 0 & \omega_{3,3} & \omega_{3,4} \\ 0 & 0 & \omega_{3,4} & \omega_{4,4} \end{bmatrix}.$$

The pattern of zeros in the covariance matrix of the measurement errors is not arbitrary. It says that  $W_{i,1}$  and  $V_{i,1}$  form one set of measurements, while  $W_{i,2}$  and  $V_{i,2}$  form a second set. Measurement errors may be correlated within sets, but not between sets. The two sets of data would be collected at separate times and perhaps by separate methods.

- (a) Calculate the variance-covariance matrix of the observable variables. Be careful; the measurement error terms are not all independent, and the expected value of the product is not always the product of expected values; Look at  $\Omega$  to tell. Show your work.
- (b) Write down the moment structure equations.
- (c) Are the parameters of this model identifiable? Answer Yes or No and prove your answer.

4. In this problem,  $Y_{i,1}$  is the dependent variable of primary interest, while  $Y_{i,2}$  and  $Y_{i,3}$  are instrumental variables. The point of the question is that the error terms of instrumental variables need not all be independent.

Independently for  $i = 1, \dots, n$ ,

$$\begin{aligned} Y_{i,1} &= \beta_{0,1} + \beta_{1,1}X_i + \epsilon_{i,1} \\ Y_{i,2} &= \beta_{0,2} + \beta_{1,2}X_i + \epsilon_{i,2} \\ Y_{i,3} &= \beta_{0,3} + \beta_{1,3}X_i + \epsilon_{i,3} \\ W_i &= X_i + e_i \end{aligned}$$

where

- $X_i \sim N(\mu_x, \phi)$  is a latent variable
- $e_i \sim N(0, \omega)$
- $\epsilon_i = (\epsilon_{i,1}, \epsilon_{i,2}, \epsilon_{i,3})'$
- $X_i$ ,  $e_i$  and  $\epsilon_i$  are independent of one another
- $\epsilon_i$  is multivariate normal with mean zero and covariance matrix

$$\mathbf{\Psi} = \begin{bmatrix} \psi_{1,1} & \psi_{1,2} & 0 \\ \psi_{1,2} & \psi_{2,2} & \psi_{2,3} \\ 0 & \psi_{2,3} & \psi_{3,3} \end{bmatrix}.$$

- (a) What is the parameter vector  $\theta$  for this model?
- (b) How many moment structure equations are there. You do not have to say what they are; just give a number. Don't forget the means.
- (c) Does this problem pass the test of the Counting Rule? Answer Yes or No.
- (d) Calculate the variance-covariance matrix of the observable variables. Remember that some covariances between errors are non-zero. Show your work.
- (e) Solving the complete set of moment structure equations can be done<sup>1</sup> but it's a big chore. The primary interest is in the parameter  $\beta_{1,1}$ . Show that just this parameter is identifiable.

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<sup>1</sup>Even the intercepts are identifiable from the mean vector  $\mu$ , because there is no measurement bias term in this model. That's unrealistic, of course.