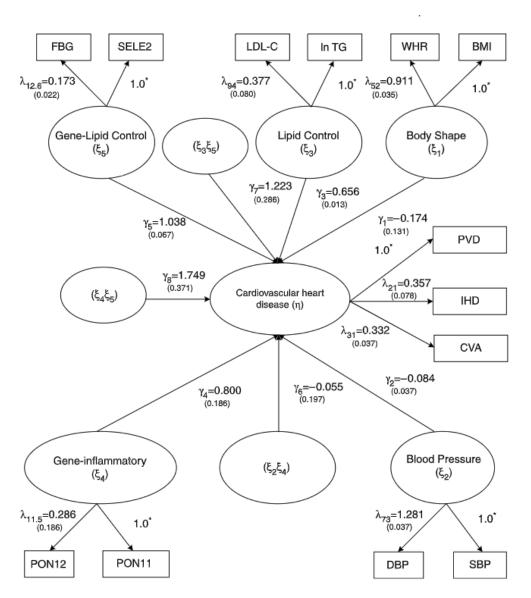
STAT 5020: Topics in Multivariate Analysis

Assignment 1 (Due date: 1-Mar-2023)

Academic year 22/23, 2nd term

1. The path diagram of a structural equation model (SEM) is presented in the following figure, along with the estimates of important parameters. Assume that the response variables are continuous.



- a. Write down the explicit form of the model.
- b. Interpret the model in terms of the variables provided in the rectangles and ellipses.
- c. Is the model identified? Why?
- d. List and interpret the fixed and unknown parameters in the model.
- e. Discuss all possible submodels of model (a).
- f. Discuss the differences between SEMs and conventional regression models.

2. A non-linear SEM is defined as follows: for i = 1, ..., n,

$$y_{i1} = \mu_1 + a_1 * c_i + \eta_{i1} + \epsilon_{i1},$$

$$y_{i2} = \mu_2 + a_2 * c_i + \lambda_{21} * \eta_{i1} + \epsilon_{i2},$$

$$y_{i3} = \mu_3 + a_3 * c_i + \eta_{i2} + \epsilon_{i3},$$

$$y_{i4} = \mu_4 + a_4 * c_i + \lambda_{42} * \eta_{i2} + \epsilon_{i4},$$

$$y_{i5} = \mu_5 + a_5 * c_i + \xi_{1i} + \epsilon_{i5},$$

$$y_{i6} = \mu_5 + a_6 * c_i + \lambda_{63} * \xi_{i1} + \epsilon_{i6},$$

$$y_{i7} = \mu_7 + a_7 * c_i + \lambda_{73} * \xi_{i1} + \epsilon_{i7},$$

$$y_{i8} = \mu_8 + a_8 * c_i + \xi_{i2} + \epsilon_{i8},$$

$$y_{i9} = \mu_9 + a_9 * c_i + \lambda_{94} * \xi_{i2} + \epsilon_{i9},$$

$$y_{i,10} = \mu_{10} + a_{10} * c_i + \lambda_{10,4} * \xi_{i2} + \epsilon_{i,10},$$

$$\eta_{i1} = b_1 * d_{i1} + b_2 * d_{i2} + b_3 * d_{i3} + \gamma_1 * \xi_{i1} + \gamma_2 * \xi_{i2} + \gamma_3 * \xi_{i1}^2 + \delta_{i1},$$

$$\eta_{i2} = b_4 * d_{i1} + b_5 * d_{i2} + b_6 * d_{i3} + \gamma_4 * \eta_{i1} + \gamma_5 * \xi_{i1} + \gamma_6 * \xi_{i2} + \gamma_7 * \xi_{i1} * \xi_{i2} + \gamma_8 * \xi_{i2}^2 + \delta_{i2},$$

where

- $\xi_i = (\xi_{i1}, \xi_{i2})^T \sim N(0, \Phi)$ with Φ being a general covariance matrix,
- $\epsilon_i = (\epsilon_{i1}, \dots, \epsilon_{i,10})^T \sim N(0, \Psi_{\epsilon}) \text{ with } \Psi_{\epsilon} = diag(\psi_{\epsilon 1}, \dots, \psi_{\epsilon 10}),$
- $\boldsymbol{\delta}_i = (\delta_{i1}, \delta_{i2})^T \sim N(0, \boldsymbol{\Psi}_{\delta}) \text{ with } \boldsymbol{\Psi}_{\delta} = diag(\psi_{\delta 1}, \psi_{\delta 2}),$
- a. Write the model in a matrix form and draw the corresponding path diagram.
- b. In a Bayesian analysis, what prior distributions do we usually assign to the model parameters? Why?
- c. Derive the full conditional distributions in detail for the unknown parameters.