### 5 November 2020

### One categorical variable + one continuous variable

## Model I - Regression Model

$$y_i = \beta_0 + \sum_{j=1}^{m-1} \beta_{g_j} * g_{i,j} + \beta_1 * x_{i1} + \sum_{j=1}^{m-1} \beta_{1g_j} * g_{i,j} * x_{i1} + e_i$$

for  $i=1,\ldots,n$ , where  $g_{ij}=1$  if  $i^{th}$  observation is in  $j^{th}$  level and  $g_{ij}=0$  otherwise. The terms of  $g_{i,j}*x_{i1}$ , for  $j=1,\ldots,m-1$ , are called interaction terms.

#### Model II - ANCOVA Model

$$y_{ij} = \gamma_{0i} + \gamma_{1i}x_{ij,1} + e_{ij}, \qquad i = 1, \dots, m, \ j = 1, \dots, n_i$$

Write

$$\beta = \begin{pmatrix} \gamma_{01} \\ \gamma_{11} \\ \gamma_{02} \\ \gamma_{12} \\ \vdots \\ \gamma_{0m} \\ \gamma_{1m} \end{pmatrix}$$

then,

$$\underline{\boldsymbol{\mathcal{X}}}^T \underline{\boldsymbol{\mathcal{X}}} = \begin{pmatrix} \underline{\boldsymbol{\mathcal{X}}}_1^T \underline{\boldsymbol{\mathcal{X}}}_1 & \underline{\boldsymbol{\mathcal{Q}}} & \underline{\boldsymbol{\mathcal{Q}}} & \dots & \underline{\boldsymbol{\mathcal{Q}}} \\ \underline{\boldsymbol{\mathcal{Q}}} & \underline{\boldsymbol{\mathcal{X}}}_2^T \underline{\boldsymbol{\mathcal{X}}}_2 & \underline{\boldsymbol{\mathcal{Q}}} & \dots & \underline{\boldsymbol{\mathcal{Q}}} \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ \underline{\boldsymbol{\mathcal{Q}}} & \underline{\boldsymbol{\mathcal{Q}}} & \underline{\boldsymbol{\mathcal{Q}}} & \dots & \underline{\boldsymbol{\mathcal{X}}}_m^T \underline{\boldsymbol{\mathcal{X}}}_m \end{pmatrix} \qquad \underline{\boldsymbol{\mathcal{X}}}^T \underline{\boldsymbol{\mathcal{Y}}} = \begin{pmatrix} \underline{\boldsymbol{\mathcal{X}}}_1^T \underline{\boldsymbol{\mathcal{Y}}}_1 \\ \underline{\boldsymbol{\mathcal{X}}}_1^T \underline{\boldsymbol{\mathcal{Y}}}_1 \\ \underline{\boldsymbol{\mathcal{X}}}_2^T \underline{\boldsymbol{\mathcal{Y}}}_2 \\ \vdots \\ \underline{\boldsymbol{\mathcal{X}}}_m^T \underline{\boldsymbol{\mathcal{Y}}}_m \end{pmatrix}$$

 $\Rightarrow$ 

$$\hat{\beta} = \begin{pmatrix} \hat{\gamma}_{01} \\ \hat{\gamma}_{11} \\ \hat{\gamma}_{02} \\ \hat{\gamma}_{12} \\ \vdots \\ \vdots \\ \hat{\gamma}_{0m} \\ \hat{\gamma}_{1m} \end{pmatrix} = \begin{pmatrix} \left( X_1^T X_1 \right)^{-1} X_1^T Y_1 \\ \left( X_2^T X_2 \right)^{-1} X_2^T Y_2 \\ \vdots \\ \vdots \\ \left( X_m^T X_m \right)^{-1} X_m^T Y_m \end{pmatrix}$$

and

$$\hat{\sigma}^{2} = \frac{\sum_{i=1}^{m} Res. S. S.|_{i}}{\sum_{i=1}^{m} (n_{i} - 2)}$$

As

$$\beta_1 + \beta_{1g_1} = \gamma_{11}$$

$$\vdots :$$

$$\beta_1 + \beta_{1g_{m-1}} = \gamma_{1,m-1}$$

$$\beta_1 = \gamma_{1m}$$

 $H_o:\ \beta_{1g_1}=\beta_{1g_2}=\cdots=\beta_{1g_{m-1}}=0$  (in Model I) is equivalent to  $H_0:\ \gamma_{11}=\gamma_{12}=\ldots=\gamma_{1m}$  (in Model II)

For example, when m = 3, write  $H_0: \gamma_{11} = \gamma_{12} = \gamma_{13}$  as

$$H_0: \begin{pmatrix} 0 & 1 & 0 & 0 & 0 & -1 \\ 0 & 0 & 0 & 1 & 0 & -1 \end{pmatrix} \begin{pmatrix} \gamma_{01} \\ \gamma_{11} \\ \gamma_{02} \\ \gamma_{12} \\ \gamma_{03} \\ \gamma_{13} \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

Then, use the test statistic in Section 4.2.3. of Chapter 1 to handle.

Can't reject  $H_0$ , then re-write

#### Model I

$$y_i = \beta_0 + \sum_{j=1}^{m-1} \beta_{g_j} * g_{i,j} + \beta_1 * x_{i1} + e_i$$

for i = 1, ..., n, where  $g_{ij} = 1$  if  $i^{th}$  observation is in  $j^{th}$  level and  $g_{ij} = 0$  otherwise.

# Model II

$$y_{ij} = \gamma_{0i} + \beta_1 x_{1,ij} + e_{ij}, \qquad i = 1, \dots, m, \ j = 1, \dots, n_i$$

It is suggested to perform estimation or hypothesis testing based on Model I (transformed to centered model).