

## Chapter 2 Categorical variables

22 Oct

Model  $\underline{Y} = \underline{X} \underline{\beta} + \underline{e}$   $\underline{e} \stackrel{\text{Assumption}}{\sim} N(0, \sigma^2 \underline{I})$

$\hat{\underline{\beta}} = (\underline{X}^T \underline{X})^{-1} \underline{X}^T \underline{Y}$   $\hat{\underline{\beta}} \sim N(\underline{\beta}, \sigma^2 (\underline{X}^T \underline{X})^{-1})$

$\hat{\sigma}^2 = \frac{\text{Res S.S.}}{n - p'}$   $\frac{\text{Res S.S.}}{\sigma^2} \sim \chi^2(n - p')$

where Res S.S. =  $\underline{Y}^T (\underline{I} - \underline{X} (\underline{X}^T \underline{X})^{-1} \underline{X}^T) \underline{Y}$

$\Rightarrow$  C.I. of  $\underline{\beta}$ , t-test, F-test

Chapter 1

## One categorical variable (One-way ANOVA)

$\uparrow$

e.g. p.4

Fit  $y$  on group = 1, 2, 3, 4, 5

Model  $y = \beta_0 + \beta_1 x + e$   $\leftarrow$  Lack-of-fit

group	Model	$E(y)$
1	$y = \beta_0 + \beta_1 + e$	$\beta_0 + \beta_1 = \mu_1$
2	$y = \beta_0 + 2\beta_1 + e$	$\beta_0 + 2\beta_1 = \mu_2$
3	$y = \beta_0 + 3\beta_1 + e$	$\beta_0 + 3\beta_1 = \mu_3$
4	$y = \beta_0 + 4\beta_1 + e$	$\beta_0 + 4\beta_1 = \mu_4$
5	$y = \beta_0 + 5\beta_1 + e$	$\beta_0 + 5\beta_1 = \mu_5$

$\mu_2 - \mu_1 = \mu_3 - \mu_2 = \mu_4 - \mu_3 =$

$\mu_5 - \mu_4$

$\Downarrow$

It is not a general model

## General model Dummy variable (indicator variable)

group	$g_1$	$g_2$	$g_3$	$g_4$	$g_5$
1	1	0	0	0	0
2	0	1	0	0	0
3	0	0	1	0	0
4	0	0	0	1	0
5	0	0	0	0	1

group =  $i \Rightarrow g_i = 1$

$i = 1, 2, 3, 4, 5$

①

$$X = \begin{matrix} & \beta_0 & \alpha_1 & \alpha_2 & \alpha_3 & \alpha_4 & \alpha_5 \\ \begin{matrix} \vdots \\ \vdots \\ \vdots \\ \vdots \\ \vdots \end{matrix} & \begin{pmatrix} \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \end{pmatrix} & \begin{matrix} \text{group} = 1 \\ \text{group} = 2 \\ \text{group} = 3 \\ \text{group} = 4 \\ \text{group} = 5 \end{matrix} \end{matrix}$$

$$\det(X^T X) = 0$$

$X$  is singular matrix

# of levels of group = ~~5~~<sup>m</sup>

$\Rightarrow$  create ~~(5-1)~~<sup>m</sup> dummy variables

We need to add one constraint, e.g., delete the last column

Model I (Regression model)

+ ... +  $\alpha_{m-1} g_{i,m-1}$

$$y_i = \beta_0 + \alpha_1 g_{i1} + \alpha_2 g_{i2} + \alpha_3 g_{i3} + \alpha_4 g_{i4} + \epsilon_i \quad i=1, \dots, n$$

$\uparrow$

$\alpha_k$  - regression coeff. of  $g_k \quad k=1, 2, \dots, m-1$

Analysis of variance

ANOVA ~~tot~~ model

$$y_{ij} = \beta_0 + \alpha_i + \epsilon_{ij}$$

$$i=1, \dots, m$$

$$j=1, \dots, n_i$$

Group      Model

1       $E(y_{1j}) = \beta_0 + \alpha_1 + \cancel{\epsilon_{1j}} \quad \mu_1$

$$j=1, \dots, n_1$$

2       $E(y_{2j}) = \beta_0 + \alpha_2 + \cancel{\epsilon_{2j}} \quad \mu_2$

$$j=1, \dots, n_2$$

3       $E(y_{3j}) = \beta_0 + \alpha_3 + \cancel{\epsilon_{3j}} \quad \mu_3$

$$j=1, \dots, n_3$$

4       $E(y_{4j}) = \beta_0 + \alpha_4 + \cancel{\epsilon_{4j}} \quad \mu_4$

$$j=1, \dots, n_4$$

5       $E(y_{5j}) = \beta_0 + \cancel{\alpha_5} + \cancel{\epsilon_{5j}} \quad \mu_5$

$$j=1, \dots, n_5$$

$\Rightarrow$

$$y_{ij} = \mu_5 + \epsilon_{ij}$$

$$i=1, \dots, m$$

$$j=1, \dots, n_i$$

$$\beta_0 + \alpha_1 = \mu_1$$

$$\alpha_1 = \mu_1 - \mu_5$$

$$\beta_0 + \alpha_2 = \mu_2$$

$$\alpha_2 = \mu_2 - \mu_5$$

$$\beta_0 + \alpha_3 = \mu_3 \Rightarrow$$

$$\alpha_3 = \mu_3 - \mu_5$$

$$\beta_0 + \alpha_4 = \mu_4$$

$$\alpha_4 = \mu_4 - \mu_5$$

$$\beta_0 = \mu_5$$

group = 5 is a reference group

model without intercept

$$X = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix} \begin{matrix} \mu_1 \\ \mu_2 \\ \mu_3 \\ \mu_4 \\ \mu_5 \end{matrix}$$

group = 1  
group = 2  
group = 3  
group = 4  
group = 5

Model =  $y_{ij} = \mu_i + \rho_{ij}$

$i = 1, \dots, 5$   
 $j = 1, \dots, 6$

$$Y = \begin{pmatrix} y_{11} \\ y_{12} \\ y_{13} \\ y_{14} \\ y_{15} \\ y_{16} \\ y_{21} \\ y_{22} \\ y_{23} \\ y_{24} \\ y_{25} \\ y_{26} \\ y_{31} \\ y_{32} \\ y_{33} \\ y_{34} \\ y_{35} \\ y_{36} \\ y_{41} \\ y_{42} \\ y_{43} \\ y_{44} \\ y_{45} \\ y_{46} \\ y_{51} \\ y_{52} \\ y_{53} \\ y_{54} \\ y_{55} \\ y_{56} \end{pmatrix} \begin{matrix} \text{group} = 1 \\ \text{group} = 2 \\ \text{group} = 3 \\ \text{group} = 4 \\ \text{group} = 5 \end{matrix}$$

$$X^T X = \begin{pmatrix} n_1 & 0 & 0 & 0 & 0 \\ 0 & n_2 & 0 & 0 & 0 \\ 0 & 0 & n_3 & 0 & 0 \\ 0 & 0 & 0 & n_4 & 0 \\ 0 & 0 & 0 & 0 & n_5 \end{pmatrix}$$

5x5

$$= \begin{pmatrix} 551 \\ 632 \\ 595 \\ 519 \\ 563 \\ 679 \end{pmatrix} \begin{matrix} \text{group} = 1 \\ \text{group} = 2 \\ \text{group} = 3 \\ \text{group} = 4 \\ \text{group} = 5 \end{matrix}$$

$$(X^T X)^{-1} = \begin{pmatrix} \frac{1}{n_1} & 0 & 0 & 0 & 0 \\ 0 & \frac{1}{n_2} & 0 & 0 & 0 \\ 0 & 0 & \frac{1}{n_3} & 0 & 0 \\ 0 & 0 & 0 & \frac{1}{n_4} & 0 \\ 0 & 0 & 0 & 0 & \frac{1}{n_5} \end{pmatrix}$$

$$X^T Y = \begin{pmatrix} \sum_{j=1}^{n_1} y_{1j} \\ \sum_{j=1}^{n_2} y_{2j} \\ \vdots \\ \sum_{j=1}^{n_5} y_{5j} \end{pmatrix}$$

$$\hat{\beta} = (X^T X)^{-1} X^T Y$$

$$= \begin{pmatrix} \bar{y}_1 \\ \bar{y}_2 \\ \bar{y}_3 \\ \bar{y}_4 \\ \bar{y}_5 \end{pmatrix}$$

$$\hat{\beta} = \begin{pmatrix} \hat{\mu}_1 \\ \hat{\mu}_2 \\ \hat{\mu}_3 \\ \hat{\mu}_4 \\ \hat{\mu}_5 \end{pmatrix}$$

$$\hat{\mu}_i = \bar{y}_i, i = 1, \dots, m$$

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## Regression model

$$y_i = \beta_0 + \alpha_1 g_{i1} + \alpha_2 g_{i2} + \alpha_3 g_{i3} + \alpha_4 g_{i4} + \epsilon_i$$

## ANOVA model

$$y_{ij} = \mu_i + \epsilon_{ij} \quad i=1, \dots, 5$$

$$j=1, \dots, n_i$$

$$\begin{aligned} \hat{\beta}_0 + \hat{\alpha}_1 &= \hat{\mu}_1 = \bar{y}_{1\cdot} \\ \hat{\beta}_0 + \hat{\alpha}_2 &= \hat{\mu}_2 = \bar{y}_{2\cdot} \\ \hat{\beta}_0 + \hat{\alpha}_3 &= \hat{\mu}_3 = \bar{y}_{3\cdot} \\ \hat{\beta}_0 + \hat{\alpha}_4 &= \hat{\mu}_4 = \bar{y}_{4\cdot} \\ \hat{\beta}_0 &= \hat{\mu}_5 = \bar{y}_{5\cdot} \end{aligned}$$

$$\hat{\beta}_0 = \bar{y}_{5\cdot}, \quad \hat{\alpha}_1 = \bar{y}_{5\cdot} - \bar{y}_{1\cdot}, \dots$$

$$\text{Res S.S.} = \mathbf{Y}^T \mathbf{Y} - \hat{\boldsymbol{\beta}}^T \mathbf{X}^T \mathbf{Y}$$

$$= \sum_{i=1}^m \left[ \sum_{j=1}^{n_i} y_{ij}^2 \right] - (\bar{y}_1, \bar{y}_2, \dots, \bar{y}_m) \begin{pmatrix} \sum_{j=1}^{n_1} y_{1j} \\ \vdots \\ \sum_{j=1}^{n_m} y_{mj} \end{pmatrix}$$

For  $i=1$

$$\sum_{j=1}^{n_1} y_{1j}^2 - \bar{y}_{1\cdot} \left( \sum_{j=1}^{n_1} y_{1j} \right) \leftarrow n_1 \bar{y}_{1\cdot}$$

$$= \sum_{j=1}^{n_1} y_{1j}^2 - n_1 \bar{y}_{1\cdot}^2$$

$$= \sum_{j=1}^{n_1} (y_{1j} - \bar{y}_{1\cdot})^2$$

Pure Error S.S.

$$\Rightarrow \text{Res S.S.} = \sum_{i=1}^m \sum_{j=1}^{n_i} (y_{ij} - \bar{y}_{i\cdot})^2$$

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$$\frac{\sum_{i=1}^m \sum_{j=1}^{n_i} (y_{ij} - \bar{y}_{i\cdot})^2}{\sigma^2} \sim \chi^2_{(n_i-1)}$$

$$\frac{\text{Res S.S.}}{\sigma^2} \sim \chi^2_{\underbrace{\sum_{i=1}^m (n_i-1)}_{n-m}} \Rightarrow \hat{\sigma}^2 = \frac{\text{Res S.S.}}{n-m} = \frac{\sum_{i=1}^m \sum_{j=1}^{n_i} (y_{ij} - \bar{y}_{i\cdot})^2}{n-m}$$

$$\text{Var}(\hat{\beta}) = (X^T X)^{-1} \sigma^2$$

$$= \begin{pmatrix} \frac{1}{n_1} & 0 & 0 & 0 & 0 \\ 0 & \frac{1}{n_2} & 0 & 0 & 0 \\ 0 & 0 & \frac{1}{n_3} & 0 & 0 \\ 0 & 0 & 0 & \frac{1}{n_4} & 0 \\ 0 & 0 & 0 & 0 & \frac{1}{n_5} \end{pmatrix} \sigma^2$$

$$\beta = \begin{pmatrix} \mu_1 \\ \mu_2 \\ \mu_3 \\ \mu_4 \\ \mu_5 \end{pmatrix}$$

$$\text{Var}(\hat{\mu}_i) = \frac{\sigma^2}{n_i}$$

$$\Rightarrow \text{Var}(\bar{y}_i) = \frac{\sigma^2}{n_i}$$

C.I. for  $\mu_i$

$$\bar{y}_i \pm t_{\alpha/2, n-m} \hat{\sigma} \sqrt{\frac{1}{n_i}}$$

$$H_0: \mu_i = \mu_{i0}$$

$$t = \frac{\bar{y}_i - \mu_{i0}}{\hat{\sigma} / \sqrt{n_i}}$$

Fitted model:  $y$  on group = 1, 2, 3, 4, 5

$$y_i = \beta_0 + \beta_1 * \text{group}_i + e \quad i=1, \dots, n$$

1, 2, 3, 4, 5

Section 6 of Chapter 4

$H_0$ : No lack of fit

$H_1$ :

$$F = \frac{\boxed{\text{Lack of fit S.S.}} / \boxed{\text{d.f. | lack of fit S.S.}}}{\boxed{\text{Pure Error S.S.}} / \boxed{\text{d.f. | pure error S.S.}}}$$

$\frac{\sum_{i=1}^m \sum_{j=1}^{n_i} (y_{ij} - \bar{y}_i)^2}{n-m}$

Res S.S. | ANOVA

Res S.S. / fitted model - Pure Error S.S.  
 $n-2$   
 d.f. of Res S.S. / fitted model  
 - d.f. of Pure Error S.S.  
 $n-m$



$$H_0 = \mu_1 = \mu_2 = \dots = \mu_m = \mu$$

vs  $H_1$  : at least ~~one~~<sup>two</sup> of  $\mu$ 's are not equal

ANOVA model

$$y_{ij} = \mu_i + e_{ij}$$

Under  $H_0$ ,  $y_{ij} = \mu + e_{ij}$

Regression model

$$y_i = \beta_0 + \alpha_1 g_{i1} + \dots + \alpha_{m-1} g_{i,m-1} + e_i$$

Under  $H_0$ ,  $y_i = \beta_0 + e_i$

Section 4 of Chaptr. 1

$$\text{Total S.S.} = \text{Reg S.S.} + \text{Res S.S.}$$

$$\sum_{i=1}^m \sum_{j=1}^{n_i} (y_{ij} - \underbrace{\bar{y}_{..}}_{\substack{\text{overall mean} \\ \text{of } y_{ij}}})^2 = \sum_{i=1}^m \sum_{j=1}^{n_i} (y_{ij} - \bar{y}_{i.} + \bar{y}_{i.} - \bar{y}_{..})^2$$

$$= \sum_{i=1}^m \sum_{j=1}^{n_i} (y_{ij} - \bar{y}_{i.})^2 + \underbrace{\sum_{i=1}^m \sum_{j=1}^{n_i} (\bar{y}_{i.} - \bar{y}_{..})^2}_{\substack{\parallel \\ \sum_{i=1}^m n_i (\bar{y}_{i.} - \bar{y}_{..})^2}}$$

$$\frac{\text{Total S.S.}}{S^2} \sim \chi^2(n-1, \lambda)$$

$$\frac{\text{Res S.S.}}{S^2} \sim \chi^2(n-m)$$

$$\frac{\text{Reg S.S.}}{S^2} \sim \chi^2(m-1, \lambda)$$

$$\Rightarrow F = \frac{\text{Reg S.S.} / (m-1)}{\text{Res S.S.} / (n-m)} \underset{\text{under } H_0}{\sim} F(m-1, n-m)$$

$$\text{Reg S.S.} = \sum_{i=1}^m n_i (\bar{y}_{i.} - \bar{y}_{..})^2$$

$$= \sum_{i=1}^m n_i \left( \frac{T_{i.}}{n_i} - \frac{T_{..}}{N} \right)^2$$

$N = \sum_{i=1}^m n_i$   
= total # of observations

$$= \sum_{i=1}^m n_i \left( \frac{T_{i.}^2}{n_i^2} + \frac{T_{..}^2}{N^2} - 2 \frac{T_{i.}}{n_i} \frac{T_{..}}{N} \right) \quad \parallel T_{..}$$

$$= \sum_{i=1}^m \frac{T_{i.}^2}{n_i} + \frac{T_{..}^2}{N^2} \underbrace{\left( \sum_{i=1}^m n_i \right)}_N - 2 \frac{T_{..}}{N} \left( \sum_{i=1}^m n_i \times \frac{T_{i.}}{n_i} \right)$$

$$= \sum_{i=1}^m \frac{T_{i.}^2}{n_i} - \frac{T_{..}^2}{N}$$

$$H_0: \mu_1 = \mu_2 = \dots = \mu_m = \mu \iff H_0: \alpha_1 = \alpha_2 = \dots = \alpha_{m-1} = 0$$

Source of Variation of	Sum of Squares	Degrees of freedom	Mean Square	Computed f
Model	$\sum_{i=1}^m n_i (\bar{y}_{i.} - \bar{y}_{..})^2$	$m - 1$	$\frac{\sum_{i=1}^m n_i (\bar{y}_{i.} - \bar{y}_{..})^2}{m - 1}$	$\frac{(\sum_{i=1}^m n_i - m) \sum_{i=1}^m n_i (\bar{y}_{i.} - \bar{y}_{..})^2}{(m - 1) \sum_{i=1}^m \sum_{j=1}^{n_i} (y_{ij} - \bar{y}_{i.})^2}$
Error	$\sum_{i=1}^m \sum_{j=1}^{n_i} (y_{ij} - \bar{y}_{i.})^2$	$\sum_{i=1}^m n_i - m$	$\frac{\sum_{i=1}^m \sum_{j=1}^{n_i} (y_{ij} - \bar{y}_{i.})^2}{\sum_{i=1}^m n_i - m}$	
Total	$\sum_{i=1}^m \sum_{j=1}^{n_i} (y_{ij} - \bar{y}_{..})^2$	$\sum_{i=1}^m n_i - 1$		

The advantages of choosing equal sample sizes over the choice of unequal sample sizes are: 1) the  $f$  ratio is insensitive to slight departures from the assumption of equal variances for the  $m$  populations when the sample are of equal sizes; and 2) the choice of equal sample size minimizes the probability of committing a type II error.

Example

$S_i^2$  - sample variance of  $y_{ij}$  in the  $i$ th group

$X_1$  categorical variable

repeated measurements

	1	2	3	4	5	
$y_{i1}$	551	595	639	417	563	$y_{51}$
$y_{i2}$	457	580	615	449	631	$\vdots$
$\vdots$	450	508	511	517	522	$\vdots$
$\vdots$	731	583	573	438	613	$\vdots$
$\vdots$	499	633	648	415	656	$y_{56}$
$y_{i6}$	632	517	677	555	679	
Total	3320	3416	3663	2791	3664	16854
Mean	553.33	569.33	610.50	465.19	610.67	561.80

$S_T^2$  sample variance of all  $y_{ij}$

$S_{yy} = \text{sum of squares of } y_{ij} \text{ in the } i\text{th group}$

Solution

$$S_1^2 = 12,133.8667, S_2^2 = 2,302.6667, S_3^2 = 3593.5, S_4^2 = 3,318.5667, S_5^2 = 3,455.4667, S_T^2 = 7,219.8897$$

Source of Variation of	Sum of Squares	Degrees of freedom	Mean Square	Computed f
Group	85,356	4	21,339	4.30
Error	124,021	25	4,961	
Total	209,377	29		

① Reg S.S. =

Total S.S. - Res S.S.

②  $\frac{m}{i=1} \frac{T_{i.}^2}{n_i} - \frac{T_{..}^2}{n}$

$$\sum_{i=1}^m \sum_{j=1}^{n_i} (y_{ij} - \bar{y}_{i.})^2 = \sum_{i=1}^m (n_i - 1) S_i^2$$

C.V. =  $F_{0.05, 4, 25} = 2.76$

The critical value  $f_{0.05}(4, 25) = 2.76$ . Thus,  $H_0: \mu_1 = \mu_2 = \mu_3 = \mu_4 = \mu_5$  is rejected.

$$\sum_{i=1}^m \sum_{j=1}^{n_i} (y_{ij} - \bar{y}_{..})^2 = (n - 1) S_T^2 = S_{yy}$$

overall sample size

$\Rightarrow$  Reject  $H_0$

What is the reason of rejecting the null hypothesis?