

# Chapter 13

## Random effects

Stat 566  
3-31-15

①

$$y_{ij} = \mu + \tau_i + \varepsilon_{ij} \quad \begin{matrix} i=1, \dots, a \\ j=1, \dots, n \end{matrix}$$

$$E[\varepsilon_{ij}] \equiv 0$$

$$V[\varepsilon_{ij}] = \sigma^2$$

$\tau_i$ 's are random variables

$$E[\tau_i] \equiv 0$$

$$V[\tau_i] \equiv \sigma_\tau^2$$

All of the  $\varepsilon_{ij}$ 's and  $\tau_i$ 's are independent

$$H_0: \sigma_\tau^2 = 0$$

$$H_1: \sigma_\tau^2 > 0$$

Decomposition of sum of squares

②

$$SS_{TOT} = SS_{TRT} + SS_E$$

$$\sum_{i=1}^a \sum_{j=1}^n (y_{ij} - \bar{y}_{..})^2 = n \sum_{i=1}^a (\bar{y}_{i.} - \bar{y}_{..})^2 + \sum_{i=1}^a \sum_{j=1}^n (y_{ij} - \bar{y}_{i.})^2$$

$$\text{Find } E[SS_{TRT}] = n E\left[\sum_{i=1}^a (\bar{y}_{i.} - \bar{y}_{..})^2\right]$$

$$= n E\left[\sum_{i=1}^a \bar{y}_{i.}^2 - 2 \bar{y}_{..} \sum_{i=1}^a \bar{y}_{i.} + a \bar{y}_{..}^2\right]$$

$$= n E\left[\sum_{i=1}^a \bar{y}_{i.}^2 - a \bar{y}_{..}^2\right]$$

let  $E[\bar{y}_{i.}^2] = C_i$   
let  $E[\bar{y}_{..}^2] = D$

(3)

$$C_i: y_{ij} = \mu + \tau_i + \varepsilon_{ij}$$

$$\bar{y}_{i.} = \mu + \tau_i + \bar{\varepsilon}_{i.}$$

$$\bar{y}_{i.}^2 = \mu^2 + \tau_i^2 + \bar{\varepsilon}_{i.}^2 + 2\mu\tau_i + 2\mu\bar{\varepsilon}_{i.} + 2\tau_i\bar{\varepsilon}_{i.}$$

$$C_i = E[\bar{y}_{i.}^2] = \mu^2 + \underbrace{E[\tau_i^2]}_{V[\tau_i] + (E[\tau_i])^2} + \underbrace{E[\bar{\varepsilon}_{i.}^2]}_{V[\bar{\varepsilon}_{i.}] + (E[\bar{\varepsilon}_{i.}])^2} + 0 + 0 + 0$$

$$\sigma_\tau^2 + 0 \quad \frac{\sigma^2}{n} + 0$$

$$C_i = \mu^2 + \sigma_\tau^2 + \frac{\sigma^2}{n}$$

(4)

$$D: y_{ij} = \mu + \tau_i + \varepsilon_{ij}$$

$$\bar{y}_{..} = \mu + \bar{\tau}_{..} + \bar{\varepsilon}_{..}$$

$$\bar{y}_{..}^2 = \mu^2 + \bar{\tau}_{..}^2 + \bar{\varepsilon}_{..}^2 + 2\mu\bar{\tau}_{..} + 2\mu\bar{\varepsilon}_{..} + 2\bar{\tau}_{..}\bar{\varepsilon}_{..}$$

$$D = E[\bar{y}_{..}^2] = \mu^2 + E[\bar{\tau}_{..}^2] + E[\bar{\varepsilon}_{..}^2] + 0 + 0 + 0$$

$$= \mu^2 + \frac{\sigma_\tau^2}{a} + \frac{\sigma^2}{an}$$

$$\text{Now } E[SS_{TRT}] = n \left[ \sum_{i=1}^a C_i - aD \right]$$

$$= n \left[ a(\mu^2 + \sigma_\tau^2 + \frac{\sigma^2}{n}) - a(\mu^2 + \frac{\sigma_\tau^2}{a} + \frac{\sigma^2}{an}) \right]$$

$$= n \left[ (a-1)\sigma_\tau^2 + \sigma^2 \left( \frac{a}{n} - \frac{1}{n} \right) \right]$$

⑤

$$= (a-1) [n\sigma_\tau^2 + \sigma^2] = E[SS_{\text{TRT}}]$$

Define  $MS_{\text{TRT}} = \frac{SS_{\text{TRT}}}{a-1}$

Then  $E[MS_{\text{TRT}}] = n\sigma_\tau^2 + \sigma^2$

Under  $H_0: \sigma_\tau^2 = 0$ ,  $E[MS_{\text{TRT}}] = \sigma^2$

Recall: In the fixed effects model,

$$E[MS_{\text{TRT}}] = n \frac{\sum_{i=1}^a \tau_i^2}{a-1} + \sigma^2$$

Facts (require normality assumption on all  $\tau_i$ 's and  $\varepsilon_{ij}$ 's) ⑥

1.  $\frac{SSE}{\sigma^2} \sim \chi^2_{N-a}$  ( $N=na$ )

2.  $\frac{SS_{\text{TRT}}}{\sigma^2} \sim \chi^2_{a-1}$

3.  $SSE$  and  $SS_{\text{TRT}}$  are independent

$\Rightarrow$  Under  $H_0$ ,  $\frac{\frac{SS_{\text{TRT}}}{\sigma^2} / (a-1)}{\frac{SSE}{\sigma^2} / (N-a)} = \frac{MS_{\text{TRT}}}{MS_E} \sim F_{a-1, N-a}$

So there is no difference in the F test for  
a 1-way ANOVA with fixed vs. random effects.

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(7)

Suppose  $H_0$  is rejected and you wish to estimate  $\sigma_z^2$

$$E[MS_{\text{Treat}}] = n\sigma_z^2 + \sigma^2$$

$$E[MS_E] = \sigma^2$$

$$E\left[\frac{MS_{\text{Treat}} - MS_E}{n}\right] = \sigma_z^2$$

So an unbiased estimator of  $\sigma_z^2$  is  $\hat{\sigma}_z^2 = \frac{MS_{\text{Treat}} - MS_E}{n}$

$\hat{\sigma}_z^2$  doesn't have a known distribution.

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(8)

Consider  $\frac{\sigma_z^2}{\sigma^2 + \sigma_z^2}$

$$y_{ij} = \mu + \tau_i + \varepsilon_{ij}$$

$$V(y_{ij}) = \sigma_z^2 + \sigma^2$$

This is the proportion of the variance of  $y_{ij}$   
that is explained by the factor.

We know: Under  $H_0$ ,  $\frac{SS_{\text{Treat}}}{\sigma^2} \sim \chi_{a-1}^2$

Under either  $H_0$  or  $H_1$ ,  $\frac{SS_E}{\sigma^2} \sim \chi_{N-a}^2$

Fact: Under  $H_1$ ,  $\frac{SS_{TBT}}{\sigma^2 + n\sigma_e^2} \sim \chi^2_{a-1}$  (9)

$$\frac{\frac{SS_{TBT}}{\sigma^2 + n\sigma_e^2} / (a-1)}{\frac{SSE}{\sigma^2} / (N-a)} = \frac{MS_{TBT}}{MS_E} \frac{\sigma^2}{\sigma^2 + n\sigma_e^2} \sim F_{a-1, N-a}$$

To be finished next time.