

Unbalanced data in factorial designs

Stat 526
T-27-25

①

		Temp (°F)					
		15		70		125	
Material Type	1	130 155	74 180	34 40	80 75	70 58	
	2	159 26	136 115	45			
	3	138 160	150 139	96			

②

Look at the
cell counts

4	4	2	10
2	2	1	5
2	2	1	5
8	8	4	20

Notice that the rows are proportional
and the columns are proportional

and
$$n_{ij} = \frac{n_{i.} \cdot n_{.j}}{n_{..}}$$

This is called proportional data

(3)

You may be able to run this using the
balanced ANOVA module in the stat
package

Exact Analysis

$$\text{Model: } y_{ijk} = \mu + \tau_i + \beta_j + (\tau\beta)_{ij} + \varepsilon_{(ijk)}$$

$$i = 1, 2, 3 \quad j = 1, 2, 3 \quad k = 1, \dots, n_{ij}$$

$$\sum_i \tau_i = 0 \quad \sum_j \beta_j = 0 \quad \sum_i \tau\beta_{ij} = \sum_j \tau\beta_{ij} = 0$$

(4)

$$\begin{array}{c}
 \begin{array}{c} \text{13} \\ \text{+ 3} \end{array} \\
 \begin{array}{c} \mu \quad \tau_1 \quad \tau_2 \quad \beta_1 \quad \beta_2 \quad \tau\beta_{11} \quad \tau\beta_{12} \quad \tau\beta_{13} \end{array} \\
 \begin{array}{c} 9 \times 1 \\ 20 \times 9 \end{array}
 \end{array}
 =
 \begin{array}{c}
 \begin{array}{c} \text{13} \\ \text{+ 3} \end{array} \\
 \begin{array}{c} 130 \quad 155 \quad 74 \quad 180 \quad 34 \quad 40 \quad 88 \quad 75 \quad 70 \quad 58 \quad 159 \quad 126 \quad 136 \quad 115 \quad 45 \quad 138 \quad 160 \quad 130 \quad 139 \quad 96 \end{array} \\
 \begin{array}{c} 20 \times 1 \\ \text{y} \end{array}
 \end{array}$$

(5)

Note:

$$\begin{aligned}\hat{\tau}\beta_{13} &= -\hat{\tau}\beta_{11} - \hat{\tau}\beta_{12} \\ \hat{\tau}\beta_{23} &= -\hat{\tau}\beta_{21} - \hat{\tau}\beta_{22} \\ \hat{\tau}\beta_{31} &= -\hat{\tau}\beta_{11} - \hat{\tau}\beta_{21} \\ \hat{\tau}\beta_{32} &= -\hat{\tau}\beta_{12} - \hat{\tau}\beta_{22} \\ \hat{\tau}\beta_{33} &= -\hat{\tau}\beta_{31} - \hat{\tau}\beta_{32} \\ &= \hat{\tau}\beta_{11} + \hat{\tau}\beta_{21} + \hat{\tau}\beta_{12} + \hat{\tau}\beta_{22}\end{aligned}$$

(6)

To get an exact analysis, run this as a regression, entering the variables $Y, X_1, X_2, X_3, X_4, X_5, X_6, X_7, X_8$ as your data set.

Look at the sequential sums of squares.

$$SS_A = SS_{X_1} + SS_{X_2|X_1}$$

$$SS_{B|A} = SS_{X_3|X_1, X_2} + SS_{X_4|X_1, X_2, X_3}$$

$$SS_{AB|A, B} = SS_{X_5|X_1, X_2, X_3, X_4} + \dots$$

(7)

Approximate methods:

Imputation: In Stat 205, we considered the case of an empty cell.

If a cell has some values, but fewer than the other cells, you can use $\bar{y}_{ij}$ as the imputed value.

You must manually subtract an error δ for each imputed value.

(8)

Setting aside a data value:

Randomly set aside values from the over-populated cells.

Yates' Method: Compute the cell means + treat those as if they were the data set.

⑨

		Temp		
		15	70	125
Type	1	134.75	57.25	64
	2	142.5	125.5	45
	3	149	144.5	96

Run this as a 2-way ANOVA with interaction. There will 0 df for error.

Compute
$$\hat{\sigma}^2 = \frac{\sum_i \sum_j \sum_k (y_{ijk} - \bar{y}_{ij.})^2}{n_{..} - ab}$$

This estimates $\sigma^2 = V(y_{ijk})$

⑩

However, Yates' Method is an ANOVA on the cell means, i.e. the $\bar{y}_{ij.}$, not the original y_{ijk} .

$$V(\bar{y}_{ij.}) = \frac{\sigma^2}{n_{ij}} \quad (\text{not the same for every cell})$$

Their average is
$$\bar{V}(\bar{y}_{ij.}) = \frac{\sum_i \sum_j \frac{\sigma^2}{n_{ij}}}{ab}$$

(11)

This can be estimated by

$$\frac{\hat{\sigma}^2}{ab} \sum_i \sum_j \frac{1}{n_{ij}}$$

Yates said to use this as the manually-computed MSE, with $n_{..} - ab$ df.

(12)

Cell-means ANOVA

Source	df
A	a-1
B	b-1
AB	(a-1)(b-1)
Error	0
Total	ab-1

Then, manually compute the MSE
+ reconstruct the ANOVA table

Source	df
A	a-1
B	b-1
AB	(a-1)(b-1)
Error	$n_{..} - ab$
Total	$n_{..} - 1$