

# Stat 541 Homework #1

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February 1, 2016

1. 
$$\begin{aligned}\sum_{i=1}^a \sum_{j=1}^{n_i} (y_{ij} - \bar{y}_{..})^2 &= \sum_{i=1}^a \sum_{j=1}^{n_i} (y_{ij}^2 - 2y_{ij}\bar{y}_{..} + \bar{y}_{..}^2) \\&= \sum_{i=1}^a \left( \sum_{j=1}^{n_i} y_{ij}^2 - 2 \left( \sum_{j=1}^{n_i} y_{ij} \right) \bar{y}_{..} + \sum_{j=1}^{n_i} \bar{y}_{..}^2 \right) \\&= \sum_{i=1}^a \left( \sum_{j=1}^{n_i} y_{ij}^2 - 2y_{i.}\bar{y}_{..} + n_i\bar{y}_{..}^2 \right) \\&= \sum_{i=1}^a \sum_{j=1}^{n_i} y_{ij}^2 - 2 \left( \sum_{i=1}^a y_{i.} \right) \bar{y}_{..} + \left( \sum_{i=1}^a n_i \right) \bar{y}_{..}^2 \\&= \sum_{i=1}^a \sum_{j=1}^{n_i} y_{ij}^2 - 2y_{..}\bar{y}_{..} + N\bar{y}_{..}^2 \\&= \sum_{i=1}^a \sum_{j=1}^{n_i} y_{ij}^2 - 2y_{..}\frac{y_{..}}{N} + N \left( \frac{y_{..}}{N} \right)^2 \\&= \sum_{i=1}^a \sum_{j=1}^{n_i} y_{ij}^2 - 2\frac{y_{..}^2}{N} + \frac{y_{..}^2}{N} \\&= \sum_{i=1}^a \sum_{j=1}^{n_i} y_{ij}^2 - \frac{y_{..}^2}{N}\end{aligned}$$

2. (a) Under  $H_0$  the test statistic is  $F_0 \sim F(5, 24)$  so for each  $\alpha$  we reject  $H_0$  if:

Significance Level	Rejection Region
$\alpha = 0.01$	$F_0 \geq F_{0.01}(5, 24) = 3.90$
$\alpha = 0.05$	$F_0 \geq F_{0.05}(5, 24) = 2.62$
$\alpha = 0.10$	$F_0 \geq F_{0.10}(5, 24) = 2.10$

- (b) First we need the overall total:

$$y_{..} = \sum_i \sum_j y_{ij} = \sum_i y_{i.} = 50 + 75 + 100 + 90 + 60 + 75 = 450$$

Next, find  $SS_T$  and  $SS_{Trt}$ :

$$SS_T = \sum_i \sum_j (y_{ij} - \bar{y}_{..})^2 = \sum_i \sum_j y_{ij}^2 - \frac{y_{..}^2}{N} = 7858 - \frac{450^2}{30} = 1108$$

$$SS_{Trt} = \sum_i \sum_j (\bar{y}_{i.} - \bar{y}_{..})^2 = \sum_i \frac{y_{i.}^2}{n_i} - \frac{y_{..}^2}{N} = \frac{50^2}{5} + \frac{75^2}{5} + \frac{100^2}{5} + \frac{90^2}{5} + \frac{60^2}{5} + \frac{75^2}{5} - \frac{450^2}{30} = 340$$

Now we can fill in the ANOVA table:

Source of Variation	Sum Squares	d.f.	Mean Square	F-Ratio
Treatment	340	5	68	$F_0 = 2.125$
Error	768	24	32	
Total	1108	29		

- (c) We would reject  $H_0$ :  $\mu_1 = \mu_2 = \mu_3 = \mu_4 = \mu_5 = \mu_6$  for  $\alpha = 0.10$  only.
- (d) With all of the original observations, the treatment means were  $\bar{y}_{1.} = 10$ ,  $\bar{y}_{2.} = 15$ ,  $\bar{y}_{3.} = 20$ ,  $\bar{y}_{4.} = 18$ ,  $\bar{y}_{5.} = 12$ , and  $\bar{y}_{6.} = 15$ . Treatment 3 had the largest mean. With the one observation removed, the new mean for treatment 3 is  $\bar{y}_{3.} = 16$  which is close to the means for treatments 2 and 6 and is no longer the largest mean. Thus there is less between-treatment variability so  $MS_{Trt}$  will decrease.

3. I ran the following SAS code:

```
data catalysts;
  input catalyst concentration @@;
datalines;
1 58.2 1 57.2 1 58.4 1 55.8 1 54.9
2 56.3 2 54.5 2 57.0 2 55.3
3 50.1 3 54.2 3 55.4
4 52.9 4 49.9 4 50.0 4 51.7
;

proc glm data = catalysts;
  class catalyst;
  model concentration = catalyst;
```

Here is the resulting ANOVA table:

The GLM Procedure					
Dependent Variable: concentration					
Source	DF	Sum of Squares	Mean Square	F Value	Pr > F
Model	3	85.6758333	28.5586111	9.92	0.0014
Error	12	34.5616667	2.8801389		
Corrected Total	15	120.2375000			

4. (a) Yes, we can compute:

$$SS_E = \sum_i (n_i - 1)s_i^2$$

(b) No, we need the sample variance  $s^2$  for the whole sample so we can compute:

$$SS_{T_{rt}} = \left( \sum_i n_i - 1 \right) s^2$$

(c) No, we need  $s^2$  to find  $SS_{T_{rt}}$  as above and then we can compute:

$$SS_T = SS_{T_{rt}} + SS_E$$