

# Homework #9

$$1) S^2 = \frac{SSE}{n-(k+1)} = \frac{1107.01}{11} = 100.637$$

$$n=15, \hat{\beta}_0 = -0.92$$

$$df = 15 - 4 = 11$$

$$V(\hat{\beta}_0) = s^2 \sigma^2 \approx \frac{8.1}{10000} (100.637) = 0.081516$$

$$\sqrt{V(\hat{\beta}_0)} \approx 0.2855$$

$$H_0: \beta_0 = 0$$

$$H_a: \beta_0 < 0$$

$$t = \frac{\hat{\beta}_0 - 0}{\sqrt{V(\hat{\beta}_0)}} = \frac{-0.92}{0.2855} = -3.22$$

$$p = P(t < -3.22 \mid df=11)$$

$$\text{Using table, } 0.005 > p > 0.0025$$

Conclusion: small p-value, reject  $H_0$ .

likelihood is that  $\beta_1 < 0$ .

i.e.  $\beta_1$  is important to the model

$$b) \hat{y}(914, 65, 6) = 39.9812$$

$$t_{0.025}(11) = 2.201$$

$$\text{let } \vec{a} = \langle 1, 914, 65, 6 \rangle$$

$$\text{By above } S = \sqrt{100.637} \text{ and}$$

$$\hat{y} \pm t_{0.025}(11) S \sqrt{\vec{a}^T (X^T X)^{-1} \vec{a}}$$

$$39.9812 \pm 2.201 \sqrt{100.637} \sqrt{\frac{927662}{10000}}$$

$$39.9812 \pm 212.664$$

$$(-172.683, 252.645)$$

2.  $k=3$   $n_1=3$ ,  $\bar{y}_1=2$   
 a)  $n_2=2$ ,  $\bar{y}_2=5$   
 $n_3=3$ ,  $\bar{y}_3=6$

$$\bar{y}_0 = \frac{1}{8} (3 \cdot 2 + 2 \cdot 5 + 3 \cdot 6) = 4.25$$

$$\begin{aligned} SSA &= \sum n_i (\bar{y}_i - \bar{y}_0)^2 \\ &= 3(-2.25)^2 + 2(0.75)^2 + 3(1.75)^2 \\ &= 25.5 \end{aligned}$$

$$\begin{aligned} SSU &= \sum_{i=1}^3 \sum_{j=1}^{n_i} (x_{ij} - \bar{y}_i)^2 \\ &= (1-2)^2 + (3-2)^2 + (2-2)^2 + (6-5)^2 + (4-5)^2 \\ &\quad + (6-6)^2 + (8-6)^2 + (4-6)^2 \\ &= 12 \end{aligned}$$

b)

	SS	df	MS	Fobs	p-value
Among	25.5	2	12.75	5.3125	0.0579
within	12	5	2.4		by R
total	37.5	7			

c) Yes.  $p = 0.0579 < 0.10$   
 At the significance level, we reject  $H_0$ .

d) This is silly Q because I forgot to say "Use Table 7"  
 By Table 7,  $0.100 > P(F > 5.3125) > 0.050$



$$3. a) \hat{\mu}_0 = \frac{n_1 \hat{\mu}_1 + n_2 \hat{\mu}_2}{n_1 + n_2} = \frac{50 \cdot 26 + 80 \cdot 25}{130} = 25.3846$$

$$\begin{aligned} SSA &= \sum_i n_i (\hat{\mu}_i - \hat{\mu}_0)^2 \\ &= 50 (26 - 25.3846)^2 + 80 (25 - 25.3846)^2 \\ &= 30.7692 \end{aligned}$$

$$\begin{aligned} SSU &= \sum_i (n_i - 1) S_i^2 \\ &= 49 (3.5)^2 + 79 (2)^2 \\ &= 1311.25 \end{aligned}$$

b)	SS	df	MS	$F_k$	$P(>F)$
Among	30.7692	1	30.7692	3.0359	0.08548
within	1311.25	128	10.2441		

Note  $n=130$ ,  $k=2$

At the 5% level, we can not reject the null hypothesis. That is, there is no statistical evidence that the prep course helped student outcomes on the SAT.

c)  $df = 128 \gg 30$  We can use z-scores.

$$(\bar{\mu}_1 - \bar{\mu}_2) \pm 1.960 \sqrt{\frac{S_1^2}{n_1} + \frac{S_2^2}{n_2}}$$

$$1 \pm 1.960 \sqrt{\frac{(3.4)^2}{50} + \frac{3^2}{80}}$$

$$1 \pm 1.17191 \text{ or } (-0.1719, 2.1719)$$

The true difference in means between the two cohorts is between  $-0.1719$  and  $2.1719$ , parts with 95% likelihood.

d)  $\mu_1 - \mu_2 = 0$  lies within this interval.  
This is equivalent to rejecting the 2-sided test at the 5% level.

4a)

```
> scores=c(45,47,67,61,56,72,39,39,43,46,59,27,60,42,55,44,30,18,56,61,60,47,69,42)
> treat=c(rep("A",6),rep("B",6),rep("C",6),rep("D",6))
> table=data.frame(scores,treat)
> results=aov(scores~treat, data=table)
> summary(results)
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
treat	3	1380	460.2	3.251	0.0434 *
Residuals	20	2831	141.6		

4b) At the 5% significance level, the p-value of 0.0434 would indicate that the mean starch level in the crop is not the same throughout all four plots. (At the 1% level, we would conclude the opposite.)

$$5. L(\text{all } Y_{ij} | \hat{\mu}_1, \hat{\mu}_2, \dots, \hat{\mu}_k, S_e^2)$$

$$= \prod_{i=1}^k \prod_{j=1}^{n_i} \frac{1}{\sqrt{2\pi S_e^2}} \exp \left( -\frac{(Y_{ij} - \hat{\mu}_i)^2}{2S_e^2} \right)$$

$$= \left( \frac{1}{2\pi S_e^2} \right)^{n/2} \exp \left( -\frac{1}{2S_e^2} \sum_{i=1}^k \sum_{j=1}^{n_i} (Y_{ij} - \hat{\mu}_i)^2 \right)$$

define this as the stat T.

$$= \left( \frac{1}{2\pi S_e^2} \right)^{n/2} \exp \left( -\frac{T}{2S_e^2} \right)$$

easier to differentiate  $\ln L$ .

$$\ln L = -\frac{n}{2} \ln(2\pi S_e^2) - \frac{T}{2S_e^2}$$

$$= -\frac{n}{2} \ln 2\pi - \frac{n}{2} \ln S_e^2 - \frac{T}{2S_e^2}$$

$$\text{Then } \frac{d \ln L}{d(S_e^2)} = -\frac{n}{2S_e^2} + \frac{T}{2(S_e^2)^2}$$

$$\frac{d \ln L}{d(S_e^2)} = 0 \iff -\frac{n}{2S_e^2} + \frac{T}{2(S_e^2)^2} = 0$$

$$-nS_e^2 + T = 0$$

$$\Rightarrow S_e^2 = \frac{T}{n} = \frac{1}{n} \sum_{i=1}^k \sum_{j=1}^{n_i} (Y_{ij} - \hat{\mu}_i)^2$$

as stated in class.



6. a) Recall for  $\text{Gamma}$ ,  $\Gamma(z) = \int_0^{\infty} x^{z-1} e^{-x} dx$ .

We have  $\int_0^{\infty} x^{\frac{n}{2}-2} e^{-x/2} dx = I$

Let  $y = x/2 \rightarrow dy = dx/2$  and  $2dy = dx$ .

$$\begin{aligned} \text{So } I &= \int_0^{\infty} (2y)^{\frac{n}{2}-2} e^{-y} \cdot 2dy \\ &= 2^{\frac{n}{2}-1} \int_0^{\infty} y^{\frac{n}{2}-2} e^{-y} dy \\ &= 2^{\frac{n}{2}-1} \int_0^{\infty} y^{z-1} e^{-y} dy \text{ where } z = \frac{n}{2}-1. \\ &= 2^{\frac{n}{2}-1} \Gamma\left(\frac{n}{2}-1\right). \end{aligned}$$

b) For  $X \sim \chi^2(n) = \Gamma\left(\frac{n}{2}, 2\right)$ , the associated PDF is,

$$f(x) = \frac{x^{\frac{n}{2}-1} e^{-x/2}}{\Gamma\left(\frac{n}{2}\right) 2^{n/2}}.$$

$$\begin{aligned} \text{Then } E\left[\frac{1}{x}\right] &= \int_0^{\infty} \frac{1}{x} f(x) dx \\ &= \int_0^{\infty} \frac{x^{\frac{n}{2}-2} e^{-x/2}}{\Gamma\left(\frac{n}{2}\right) 2^{n/2}} dx \\ &= \frac{1}{\Gamma\left(\frac{n}{2}\right) 2^{n/2}} \int_0^{\infty} x^{\frac{n}{2}-2} e^{-x/2} dx \end{aligned}$$



$$\begin{aligned}
 \text{by (a)} &= \frac{2^{n/2-1} \Gamma(\frac{n}{2}-1)}{\Gamma(n/2) 2^{n/2}} = \frac{1}{2} \cdot \frac{(\frac{n}{2}-1-1)!}{(\frac{n}{2}-1)!} \\
 &= \frac{1}{2} \cdot \frac{1}{\frac{n}{2}-1} = \frac{1}{n-2}.
 \end{aligned}$$

$$\begin{aligned}
 \text{c) } E\left[\frac{X_1/m}{X_2/n}\right] &= E\left[\frac{1}{m} X_1 \frac{1}{X_2}\right] \\
 &= \frac{n}{m} E(X_1) E\left(\frac{1}{X_2}\right) \text{ by independence} \\
 &= \frac{n}{m} \overset{m}{\uparrow} \cdot \frac{1}{n-2} \text{ by (b)} \\
 &\quad \text{by properties of } \chi^2(m) \\
 &= \frac{n}{n-2}.
 \end{aligned}$$