

**The University of Texas at Austin**  
**HOMEWORK ASSIGNMENT 4**  
*Predictive Analytics*

February 21, 2026

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**Instructions:** Provide your complete solution to the following problems. Final answers only, without appropriate justification, will receive zero points even if correct.

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**$F$  —distribution.**

**Problem 4.1.** (5 points) *Source: Ramachandran and Tsokos: Mathematical Statistics with Applications in R.* Let  $S_1^2$  denote the sample variance for a random sample of size 10 from a normal population I and let  $S_2^2$  denote the sample variance for a random sample of size 8 from a normal population II. The variance of population I is assumed to be three times the variance of population II. Assume that the two samples are **independent**. Find two numbers  $a$  and  $b$  such that

$$\mathbb{P}[S_1^2/S_2^2 \leq a] = 0.05 \quad \text{and} \quad \mathbb{P}[S_1^2/S_2^2 \geq b] = 0.05$$

**Solution.** Using obvious notation, we know from the problem that  $\sigma_1^2 = 3\sigma_2^2$ . From the theorem we did in class, we know that

$$\frac{S_1^2/\sigma_1^2}{S_2^2/\sigma_2^2} = \frac{S_1^2/3\sigma_2^2}{S_2^2/\sigma_2^2} = \frac{S_1^2}{3S_2^2}$$

has the  $F$  —distribution with  $\nu_1 - 1 = 10 - 1 = 9$  numerator degrees of freedom and  $\nu_2 - 1 = 8 - 1 = 7$  denominator degrees of freedom. It is, thus, convenient to rewrite the specified probabilities as

$$\mathbb{P}[S_1^2/3S_2^2 \leq a/3] = 0.05 \quad \text{and} \quad \mathbb{P}[S_1^2/3S_2^2 \geq b/3] = 0.05$$

Using **R**, we get

$$a = 3 * qf(0.05, 9, 7) = 0.9110937 \quad \text{and} \quad b = 3 * qf(0.95, 9, 7) = 11.03002.$$

**Problem 4.2.** (10 points) *Source: An old CAS exam problem.* A sample of size 20 is fitted to a linear regression model of the form

$$y_i = \beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2} + \beta_3 x_{i3} + \beta_4 x_{i4} + \beta_5 x_{i5} + \varepsilon_i.$$

The resulting  $F$  —ratio used to test the hypothesis

$$H_0 : \beta_1 = \beta_2 = \beta_3 = \beta_4 = \beta_5 = 0$$

is equal to 21. Determine  $R^2$ .

**Solution.** In general, for a multiple linear regression, we have that

$$\begin{aligned}
F_{p,n-p-1} &= \frac{(\text{TSS} - \text{RSS})/p}{\text{RSS}/(n-p-1)} = \frac{n-p-1}{p} \times \frac{\text{TSS} - \text{RSS}}{\text{RSS}} \\
&= \frac{n-p-1}{p} \times \frac{1 - \frac{\text{RSS}}{\text{TSS}}}{\frac{\text{RSS}}{\text{TSS}}} = \frac{n-p-1}{p} \times \frac{R^2}{1 - R^2}.
\end{aligned}$$

In this problem,

$$21 = \frac{20 - 5 - 1}{5} \times \frac{R^2}{1 - R^2} \Rightarrow 105(1 - R^2) = 14R^2 \Rightarrow R^2 = \frac{105}{119} = \frac{15}{17}.$$

**Problem 4.3.** In a simple linear regression fit on 16 observations, you obtain the point estimate of the slope parameter to be  $\hat{\beta}_1 = 3$ . The standard error of  $\hat{\beta}_1$  is estimated at 1.5.

- (10 points) Show that, in our usual notation,

$$TSS - RSS = \frac{(\sum (x_i - \bar{x})(y_i - \bar{y}))^2}{\sum (x_i - \bar{x})^2}.$$

- (10 points) Prove that for the simple linear regression, the  $F$  –statistic can be obtained as the square of the  $t$  –statistic for the slope.
- (5 points) Provide the value of the  $F$  –statistic.
- (10 points) Provide the value of the coefficient of determination  $R^2$ .

**Solution.**

- By definition,

$$\begin{aligned}
TSS - RSS &= \sum (y_i - \bar{y})^2 - \sum (y_i - \hat{y}_i)^2 \\
&= \sum (y_i - \bar{y})^2 - \sum (y_i - \hat{\beta}_0 - \hat{\beta}_1 x_i)^2 \\
&= \sum (y_i - \bar{y})^2 - \sum (y_i - \bar{y} + \hat{\beta}_1 \bar{x} - \hat{\beta}_1 x_i)^2 \\
&= \sum (y_i - \bar{y})^2 - \sum ((y_i - \bar{y}) - \hat{\beta}_1 (x_i - \bar{x}))^2 \\
&= \sum (y_i - \bar{y})^2 - \sum (y_i - \bar{y})^2 + 2\hat{\beta}_1 \sum (x_i - \bar{x})(y_i - \bar{y}) - (\hat{\beta}_1)^2 \sum (x_i - \bar{x})^2 \\
&= 2\hat{\beta}_1 \sum (x_i - \bar{x})(y_i - \bar{y}) - (\hat{\beta}_1)^2 \sum (x_i - \bar{x})^2
\end{aligned}$$

By the least-squares fit, we know that

$$\hat{\beta}_1 = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sum (x_i - \bar{x})^2}.$$

So,

$$\begin{aligned}
TSS - RSS &= 2 \left( \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sum (x_i - \bar{x})^2} \right) \sum (x_i - \bar{x})(y_i - \bar{y}) - \left( \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sum (x_i - \bar{x})^2} \right)^2 \sum (x_i - \bar{x})^2 \\
&= \frac{(\sum (x_i - \bar{x})(y_i - \bar{y}))^2}{\sum (x_i - \bar{x})^2}.
\end{aligned}$$

- See, e.g., slides <https://mcudina.github.io/page/M339G/slides/ch3-simple-linear-regression.pdf> to verify that the expression for the standard error of  $\hat{\beta}_1$  is

$$SE(\hat{\beta}_1) = \sqrt{\frac{\sigma^2}{\sum (x_i - \bar{x})^2}}.$$

Our estimate for  $\sigma^2$  is  $RSS/(n-2)$  so that the above becomes

$$SE(\hat{\beta}_1) = \sqrt{\frac{RSS}{(n-2) \sum (x_i - \bar{x})^2}}.$$

Hence, under the null, the  $t$ –statistic for  $\hat{\beta}_1$  is

$$\frac{\frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sum (x_i - \bar{x})^2}}{\sqrt{\frac{RSS}{(n-2) \sum (x_i - \bar{x})^2}}}$$

So, the square of the  $t$ –statistic for  $\hat{\beta}_1$  equals

$$\frac{\frac{(\sum (x_i - \bar{x})(y_i - \bar{y}))^2}{(\sum (x_i - \bar{x})^2)^2}}{\frac{RSS}{(n-2) \sum (x_i - \bar{x})^2}} = \frac{\frac{(\sum (x_i - \bar{x})(y_i - \bar{y}))^2}{\sum (x_i - \bar{x})^2}}{\frac{RSS}{n-2}}$$

By definition (see, e.g., slides <https://mcudina.github.io/page/M339G/slides/ch3-mlr-contd.pdf>)

$$F_{1,n-2} = \frac{TSS - RSS}{(RSS)/(n-2)}$$

What remains is to invoke part (i) of this problem.

- From the given values, the  $t$ –statistic is equal to  $3/1.5 = 2$ . So, the  $F$ –statistic is  $2^2 = 4$ .
- We can reuse the formula from the solution to the previous problem. In this problem,  $k = 1$ , and we have already calculated that  $F = 4$ . So,

$$4 = F = \frac{n-1-1}{1} \times \frac{R^2}{1-R^2} \Rightarrow 14R^2 = 4(1-R^2) \Rightarrow R^2 = \frac{2}{9}.$$