

Math 484 Quiz 5 fall 2016

Name_____

1) Suppose that the regression model is $Y_i = 10 + 2X_{1i} + \beta X_{2i} + e_i$ for i = 1, ..., n where the e_i are iid random variables. The least squares criterion is $Q(\eta) = \sum_{i=1}^{n} (Y_i - 10 - 2X_{1i} - \eta X_{2i})^2$.

a) What is
$$E(Y_i)$$
?
$$10 + 2 \times 11 + 3 \times 21$$

b) Find the least squares estimator $\hat{\beta}$ of β by setting the first derivative $\frac{d}{d\eta}Q(\eta)$ equal to zero.

$$\frac{dQ}{dm} = 2\sum_{i=1}^{n} (Y_{i} = 10 - 2X_{1i} - mX_{2i}) (-X_{2i}) \stackrel{\text{set}}{=} 0\Gamma$$

$$2m\sum_{i=1}^{n} X_{2i}^{2} = 2\sum_{i=1}^{n} X_{2i} (Y_{i} = 10 - 2X_{1i})$$
or
$$3 = \sum_{i=1}^{n} X_{2i} (Y_{i} = 10 - 2X_{1i})$$

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c) Show that your $\hat{\beta}$ is the global minimizer of the least squares criterion Q by showing that the second derivative $\frac{d^2}{d\eta^2}Q(\eta) > 0$ for all values of η .

$$\frac{d^2Q}{dm^2} = \frac{d}{dm} 2\eta = \frac{1}{2} \frac{1}{2} = \frac{1}{2} \frac{1}{2} \frac{1}{2} \frac{1}{2} \frac{1}{2} \frac{1}{2} = \frac{1}{2} \frac{1}{2} \frac{1}{2} \frac{1}{2} \frac{1}{2} = \frac{1}{2} \frac{$$

$$\frac{1}{1+2}\left(-\frac{1}{2n}\sum_{i=1}^{n}\sum_{j=1}^{n}\sum_{j=1}^{n}\sum_{j=1}^{n}\sum_{i=1}^{n}\sum_{j=1}^{n}\sum_{i=1}^{n}\sum_{j=1}^{n}\sum_{j=1}^{n}\sum_{i=1}^{n}\sum_{j=1}^{n}$$

no intercept mode!

Coefficient Estimates

Label Estimate Std. Error t-value p-value X 0.585616 0.0185600 31.553 0.0000

Sigma hat: 6.10219 Number of cases: 202 Degrees of freedom: 201

Summary Analysis of Variance Table

 Source
 df
 SS
 MS
 F
 p-value

 Regression
 1
 37071.5
 37071.5
 995.56
 0.0000

 Residual
 201
 7484.59
 37.2367

- 2) The above output is for a Cook and Weisberg (1999) Arc data set on 102 male and 100 female athletes collected at the Australian Institute of Sport. The response Y is the athlete's body fat. The variable X is BMI = body mass index $= weight/(height)^2$. Response and residual plots as well as a scatterplot of the data suggest that regression through the origin is a reasonable model.
 - a) Predict Y if X = 17.

$$\hat{q} = 0.585616(17) = [9.9555]$$

b) Predict Y if X = 34.

$$\hat{Y} = 0.585616(34) = 19.9109$$

c) Perform the t test for $\beta_1 = 0$.

reject Ho, X = BMF is needed in the MLR model for Y = body fat

d) Perform the no intercept Anova F test.

reject to there is an MLR relationship between y = body fort and X = BMI.

