

**Homework #9 – Due Monday, Dec. 15**  
COR1-GB.1305 – Statistics and Data Analysis

**Problem 1**

The file **Magazine.CSV** contains data on advertising costs and characteristics of magazines. The response variable is **PageCost**, which represents the cost of a full- page color ad in the magazine. **Circ** is the circulation of the magazine (in thousands), **MedIncome** is the median income of the readers, and **%Male** is the percentage of the readers who are male. The square root of the circulation is given in **SqrtCirc**.

- (a) Run a multiple regression of **PageCost** on **Circ**, **MedIncome** and **%Male**. Before running it, click on **Graphs**, and check the box for **Residuals plots: Four in one**. Note that the residuals versus fit plot shows structure: a generally upward-sloping pattern, with three outliers at the right dragging things down. Identify the Magazines corresponding to the three outliers (all of which have a very large circulation).
- (b) To investigate further, generate a scatterplot of **PageCost** versus **Circ**. Note that the plot is “bunched up” at the left, and “stretched out” at the right, and also a bit curved. In what way do the points identified as outliers in (a) deviate from the pattern in the plot here?
- (c) To try to improve the linear relationship, let’s try working with the square root of Circulation (**SqrtCirc**) rather than the circulation itself. Plot **PageCost** versus **SqrtCirc**. Based on the plot, explain why it seems more appropriate to use **SqrtCirc** as an explanatory variable in a linear regression rather than **Circ**.
- (d) Now, run a multiple regression of **PageCost** on **SqrtCirc**, **MedIncome** and **%Male**. Plot the residuals versus fitted values. Does it look better than in (a)? Which coefficients in the regression are statistically significant? Based on the  $p$ -values for the regression coefficients, which variables seem to be useless for predicting **PageCost**?
- (e) The  $F$ -statistic and corresponding  $p$ -value in the Analysis of Variance part of the output provides a test of the null hypothesis that the regression is useless for predicting  $Y$ , i.e., that all regression parameters besides the intercept are zero. Based on the  $p$ -value, does the regression seem to be useful for predicting  $Y$ ? Does this mean that all variables are useful? (Remember your answer to part (d)).
- (f) Re-run the regression for **PageCost**, this time with just the two explanatory variables **SqrtCirc** and **MedIncome**. Are the coefficients of both variables statistically significant? Get a 95% confidence interval for the mean page cost of a magazine with a **SqrtCirc** of 100, and a median income of \$40,000. To do this, after running the regression click on *Stat*  $\Rightarrow$  *Regression*  $\Rightarrow$  *Regression*  $\Rightarrow$  *Preict*. Then, enter 100 in the first line under **SqrtCirc** and enter 40000 in the first line under **MedIncome**. Did the  $R^2$  go down by much compared to the regression in (d)? Is the  $F$ -statistic still significant? What does this suggest about the deletion of the **%Male** variable?
- (g) Run a simple regression and scatterplot of **PageCost** versus **MedIncome**. Is the coefficient of **MedIncome** now statistically significant? Why is this puzzling in view of the regression output

in (f)? Remember, however, that the meaning and interpretation of the coefficient of a given variable depend on what other variables are included in the model.

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## Problem 2

Consider `DiamondPrices.CSV`, which you already studied in the context of simple regression. `Clarity`, `Code` and `Color` are categorical variables, which are numerically coded in `ClarityCode`, `ColorCode` and `CutCode`. (For example, Good, Very Good and Ideal cuts receive `Cut Codes` of 1, 2, 3, respectively.) Even though these are ordinal/categorical variables, please enter them in Minitab as “Continuous Predictors”, since we will treat them as numerical variables.

- (a) Run a multiple regression of `Price` on `ClarityCode` and `CutCode`. Based on the output, do these explanatory variables seem useful for predicting `Price`?
- (b) Run a multiple regression of `Price` on `ClarityCode`, `ColorCode` and `CutCode`. In what way do  $p$ -values for the individual coefficients and the  $F$ -statistic seem to provide contradictory evidence on whether any of these variables is helpful for predicting `Price` in the given model? (As we discussed in class, the  $F$ -statistic is the best place to go to make a decision on this question.) Does the  $R^2$  in this regression suggest that these three variables by themselves are very useful for predicting `Price`?
- (c) Next, run a multiple regression of `Price` on `Carats`, `ClarityCode`, `ColorCode` and `CutCode`. What happened to the variables that had insignificant coefficients in the regression in (b)? And what happened to the  $R^2$ ?
- (d) For the regression in part (c), do the  $p$ -values for the individual coefficients and the  $F$ -statistic have any apparent contradiction?
- (e) Using the regression model in part (c), predict the retail price of a diamond weighing 0.6 carats with a VS2 clarity (`ClarityCode` = 3), a G color (`ColorCode` = 4) and a Very Good cut (`CutCode` = 2).
- (f) Run a simple regression of `Price` on `Carats`, and note the  $R^2$ . Comparing this with the results from (c), do you think that the `ClarityCode`, `ColorCode` and `CutCode` (taken together) are worthwhile for predicting `Price`, above and beyond what can be obtained using `Carats` alone?

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