Problem Set #4

MACS 30100, Dr. Evans Due Monday, Feb. 5 at 11:30am

- 1. Multiple linear regression (4 points). For this problem, you will use the 397 observations from the Auto.csv dataset. This dataset includes 397 observations on miles per gallon (mpg), number of cylinders (cylinders), engine displacement (displacement), horsepower (horsepower), vehicle weight (weight), acceleration (acceleration), vehicle year (year), vehicle origin (origin), and vehicle name (name).
 - (a) Import the data using pandas.read_csv() function. Look for characters that seem out of place that might indicate missing values. Replace them with missing values using the na_values=... option.
 - (b) Produce a scatterplot matrix which includes all of the quantitative variables (mpg, cylinders, displacement, horsepower, weight, acceleration, year, origin). [Use the pandas scatterplot function in the code block below.]

- (c) Compute the correlation matrix for the quantitative variables (8×8) using the DataFrame.corr() method.
- (d) Estimate the following multiple linear regression model of mpg on all other quantitative variables, where u_i is an error term for each observation, using Python's statsmodels.api.OLS() function.

```
mpg_i = \beta_0 + \beta_1 cylinders_i + \beta_2 displacement_i + \beta_3 horsepower_i + \dots\beta_4 weight_i + \beta_5 acceleration_i + \beta_6 year_i + \beta_7 origin_i + u_i
```

- i. Which of the coefficients is statistically significant at the 1% level?
- ii. Which of the coefficients is NOT statistically significant at the 10% level?
- iii. Give an interpretation in words of the estimated coefficient $\hat{\beta}_6$ on $year_i$ using the estimated value of $\hat{\beta}_6$.
- (e) Looking at your scatterplot matrix from part (b), what are the three variables that look most likely to have a nonlinear relationship with mpq_i ?
 - i. Estimate a new multiple regression model by OLS in which you include squared terms on the three variables you identified as having a non-linear relationship to mpq_i as well as a squared term on $acceleration_i$.

 $^{^1{\}rm The~Auto.csv}$ dataset comes from James et al. (2017, Ch. 3) and is available at http://www-bcf.usc.edu/ gareth/ISL/data.html.

- ii. Report your adjusted R-squared statistic. Is it better or worse than the adjusted R-squared from part (d)?
- iii. What happened to the statistical significance of the $displacement_i$ variable coefficient and the coefficient on its squared term?
- iv. What happened to the statistical significance of the cylinders variable?
- (f) Using the regression model from part (e) and the .predict() function, what would be the predicted miles per gallon mpg of a car with 6 cylinders, displacement of 200, horsepower of 100, a weight of 3,100, acceleration of 15.1, model year of 1999, and origin of 1?
- 2. Classification problem: KKN by hand and in Python (3 points). The table below provides a training data set containing six observations, three predictors, and one qualitative response variable.

					Eucl. Dist. from
Obs.	X_1	X_2	X_3	Y	$X_1 = X_2 = X_3 = 0$
1	0	3	0	Red	
2	2	0	0	Red	
3	0	1	3	Red	
4	0	1	2	Green	
5	-1	0	1	Green	
6	1	1	1	Red	

Suppose we wish to use this data set to make a prediction for Y when $X_1 = X_2 = X_3 = 0$ using K-nearest neighbors.

(a) Compute the Euclidean distance between each observation and the test point $X_1 = X_2 = X_3 = 0$.

$$dist(\mathbf{p}, \mathbf{q}) = \sqrt{(q_1 - p_1)^2 + (q_2 - p_2)^2 + (q_3 - p_3)^2}$$

- (b) What is our KNN prediction with K = 1? Why?
- (c) What is our KNN prediction with K = 3? Why?
- (d) If the Bayes (optimal) decision boundary in this problem is highly non-linear, then would we expect the best value for K to be large or small? Why?
- (e) Use Python's scikit-learn library to estimate the KNN classifier of the test point $X_1 = X_2 = X_3 = 1$ with K = 2.
- 3. Multivariable logistic (logit) regression (3 points).

References

James, Gareth, Daniela Witten, Trevor Hastie, and Robert Tibshirani, An Introduction to Statistical Learning with Applications in R Springer Texts in Statistics, Springer, 2017.