Homework 4 - Linear Regression

NYC property sales + Boston data set (120 points)

Due Date: Monday Feb 22 at 11:59 pm

Instruction:

- Highly recommended: Watch this tutorial video on my YouTube channel before working on this assignment.
- This is a group-work assignment!
- You are expected to submit the .ipynb file and the exported .html.
- Only one member in each group needs to submit the assignment. It will be automatically submitted for the rest of group members.
- This is a long assignment, start early!
- You will be qualified to get full mark if you beat the following performance metrics:
 - Question 1: $RMSE_test = 0.6$
 - Question 2: $RMSE_test = 5.5$

Question 1 Linear regression: NYC property sales dataset (60 points)

In this exercise I want you to apply linear regression model to the NYC property sale data set that you cleaned in HW-2 EDA for NYC property sales dataset on Kaggle. You can also use my version of the dataset which is on the GitHub folder for HW4. Import the nyc-rolling-sales_clean.csv as a data frame and call it df. I specifically want you to do the followings:

- 1. Change the type of the feature variables as you see fit! You can use my answer key for HW2 as a reference. (5 points)
- 2. Define your target variable as target $= \log(\text{SALE PRICE})$ and add it to your data frame. Explain why this transformation would boost the performance of your linear model? (5 points)
- 3. Define your feature space (X). You can pick as many features as possible! it's your call! (5 points)
- 4. Use get_dummies(drop_first=True) function from pandas package to make the categorical variables into dummy variables. How many features you have now? wow! welcome to Machine Learning. (5 points)
- 5. Split the data into test (30%) and train set (70%) (5 points)
- 6. Use LinearRegression() model from Sklearn package to train the model. Do the followings: (15 points)
 - 1 Save the predicted values for the test set in y_hat_test. (5 points)
 - 2 Construct a data frame named log_predictions which has 3 columns: v_test, y_hat_test, resid.(5 points)
 - 3 Report the RMSE_test (RMSE in the test set) (5 points)
- 7. Estimate the RMSE_test using K-Fold Cross Validation technique (try K=5 and K=10) and name them as RMSE_CV5 and RMSE_CV10. (15 points)
- 8. Compare RMSE_CV with RMSE_test from part 3 and explain your observation? (5 points)

Question 2 Polynomial regression: Boston dataset (60 points)

In this exercise, you should work with the boston_polynomial.csv file which is available on the GitHub folder for HW4. Import the boston_polynomial.csv as a data frame and call it df_poly. I specifically want you to do the followings:

- 1. Define x= np.array(df_poly['LSTAT']) and y= np.array(df_poly['price']). Draw a scatter plot for price vs LSTAT using x and y. (5 points)
- 2. Import PolynomialFeatures class from sklearn.preprocessing. Now fit_transform your x and call it X_poly. Set polynomial **degree** = **5**. (5 points)
- 3. Split the data into test (30%) and train set (70%) (5 points)
- 4. Use LinearRegression() model from Sklearn package to train the model. Do the followings: (15 points)
 - 1 Save the predicted values for the test set in y_hat_test. (5 points)
 - 2 Construct a data frame named predictions which has 3 columns: y_test, y_hat_test, resid.(5 points)
 - 3 Report the RMSE_test (RMSE in the test set) (5 points)
- 5. Estimate the RMSE_test using K-Fold Cross Validation technique (K=5 only) and name it as RMSE_CV5. (10 points)
- 6. Use my_polynomial_regression() function from the notebook for class 7. With that function, construct a table with 3 columns: Degree (going from 1 to 10), RMSE_train and RMSE_test. (10 points)
- 7. Use the table from part 6 and plot the RMSE_test and RMSE_train against the Degree on the horizontal axis. (5 points)
- 8. What is the optimal polynomial degree based on your observations from the above table and chart in part 6 and 7 respectively. Explain your answer (5 points) This is called the **elbow method** by the way!