Homework #3

TECH-GB.2335.10: Data Science for Business

Due: Nov 10, 2021, 18:00

- Code **MUST** be written in Python.
- You have been provided with data containing attributes for customers of a mobile phone provider. Please download *Telecom_Customers_clustering.csv* from Brightspace.
- The file contains 20 attributes on over 7000 customers
- Before proceeding, import numpy (as np) and call the following: **np.random.seed(1103)**. This ensures that the randomness in K-Means will be the same for everyone and so outputs will be uniform. The number above, 1103, is arbitrary, use the same number.
- Please submit a single Jupyter notebook, including any code, text and graphs.

Problem 1 (40 points) Data loading and transformation

- i Load the data using pandas.read_csv() into a DataFrame
- ii Split your data into two dataframes, train and test, as we did in class (using sklearn.train_test_split) with 80% of the data as train.
- iii There are 3 numerical features: <u>tenure</u>, <u>MonthlyCharges</u>, <u>TotalCharges</u>. For these features, plot histograms of the training data. Your histograms must:
 - Use 25 bins
 - Have a grid
 - Include the feature name in the title
 - Label the Y-axis (determine what the Y-axis represents if you're not sure)
- iv For 2 of the numerical features, <u>tenure</u>, <u>MonthlyCharges</u>, calculate the mean, μ , and the standard deviation, σ , from the training data only
- v Standardize <u>tenure</u>, <u>MonthlyCharges</u> in both your train and test dataframes by subtracting corresponding μ and dividing by the corresponding σ (as computed from the training data). <u>Do not overwrite the original columns</u>. <u>Instead, create</u> two new columns: **tenure_std**, <u>MonthlyCharges_std</u>

- vi Instantiate an object of the *OneHotEncoder* class from *sklearn*, with *drop='first'*. Call *fit* on this object using only one column, *InternetService* from the training data.
 - a Using this trained object, call *transform* for the same column from both the training and test datasets and append these newly created columns to the corresponding datafames with column names obtained from the *OneHotEncoder*
 - b Why did we use drop='first'?

Problem 2 (30 points) Clustering

- i Using the standardized <u>tenure_std</u>, <u>MonthlyCharges_std</u> columns in your training data, cluster your data <u>using the KMeans class from scikit-learn</u>.
 - Experiment with number of clusters = 3, 4, 5 and 6.
 - Each time you cluster, call *predict* on your trained object (with the same two columns from your in-sample data) → this gives you cluster membership for each row. Assign these cluster values to a column called *cluster* in your train dataframe.
 - Plot a scatter plot with color-coded clusters and centroids, as we did previously in class.
 - Code to plot your clusters is below. You can pass in the variable *cluster_centers_* from your trained KMeans object, as is, for the argument centroids. **Include your 4 plots**

- ii Looking at these 4 plots, pick one that you think segments your customers into useful groups, report the choice you made. Give your segments names (or a brief description) that you think describe these segments.
- iii Now, using the choice you made for the number of clusters, and **without re-fitting** your KMeans object, call *predict()* on your test dataframe and add a *cluster* column to the dataframe and create a single plot. **Include your plot**.
- iv We fixed the random seed so results are consistent. What part of the K-Means algorithm is random?

Problem 3 (30 points) Instantiate a *sklearn.linear_model.LinearRegression()* object as we did in class and fit it with the feature *tenure* and the one hot encoded features created from the *InternetService* column. Use *MonthlyCharges* as your target variable.

Use the **training data only** for the fit!

- i What is the intercept of your linear regression
- ii What is the coefficient on the *tenure* feature?
- iii How many one hot encoded columns did you obtain from the *InternetService* column and what are the coefficients?
- iv Call predict() on your trained object with the corresponding feature columns passed in from the **test data**. Calculate r-squared using $sklearn.metrics.r2_score()$ to compare these predicted values to the target variable from your test data set.
 - a Report the r-squared value.
 - b Is this value low or high, in your opinion?