Assignment No. 6

**EECS 658** 

Introduction to Machine Learning

Due: 11:59 PM, Thursday, November 10, 2022 Submit deliverables in a single zip file to Canvas

Name of the zip file: FirstnameLastname\_Assignment6 (with your first and last name) Name of the Assignment folder within the zip file: FirstnameLastname Assignment6

## Deliverables:

- 1. Copy of Rubric6.docx with your name and ID filled out (do not submit a PDF)
- 2. Python source code.
- 3. Screen print showing the successful execution of your Python code. (Copy and paste the output from the Python console screen to a Word document and PDF it).
- 4. Answer to Part 1, Question 1.
- 5. Answer to Part 2, Question 2a.
- 6. Answer to Part 2, Question 2b.

## Assignment:

- For both parts use the entire iris data set. We don't need to do training and test sets because this is Unsupervised ML. In both parts you will cluster the data, to see if it clusters into 3 classes or not. Then, you will use the predict() function with the clusters to see how well the k-means and GMM clustered the data. Maybe you will find that there should be more or less than 3 species of iris, based on the data Fisher collected.
- Use the scikit-learn libraries I referenced in class with the default parameters unless otherwise specified below.
- Print out labels between the outputs below so it is clear what you are displaying. Part 1: k-Means Clustering
- Run the k-means algorithm for k = 1 through 20 and plot the reconstruction error vs. k. You will need to figure out how to plot something in Python. (Hint: Look at PlottingCode.py in Assignment 6 module).
- Find the "elbow" of the curve manually. We will call that the elbow\_k. (10 points extra credit, if you can find a way to find it algorithmically).
- Now use the predict() method and the clusters for k=elbow\_k to classify the entire iris data set.
- Print out the confusion matrix and accuracy.
  - When you examine the knee of the curve for k-means, you may find an elbow\_k different than 3. That is perfectly fine. There is no right way to determine k.
  - o If you selected elbow\_k=3, then you have to keep track of which cluster each class went to when you calculate the Accuracy Score. The easiest way to do that is to see what the majority class is in each cluster. Note: Look at <a href="https://stackoverflow.com/questions/45114760/how-to-plot-the-confusion-similarity-matrix-of-a-k-mean-algorithm">https://stackoverflow.com/questions/45114760/how-to-plot-the-confusion-similarity-matrix-of-a-k-mean-algorithm</a>. You can use this to match the k-mean labels (or GMM prediction) and the truth labels such that the number of true-positive predictions is maximized, essentially

- rearranging the columns so that the sum of diagonal entries is maximized. From this you can calculate accuracy score by (sum of diagonal entries)/(sum of all entries).
- O If you selected elbow\_k not equal to 3, then you cannot calculate an Accuracy Score. There are similar measures, but we didn't go over them in class, so I don't expect you to calculate them. Instead, print out a message something like this: "Cannot calculate Accuracy Score because the number of classes is not the same as the number of clusters".
- O You CAN, however, print out the Confusion Matrix, if you selected elbow\_k!=3 using the built-in Confusion Matrix function. The top 3 rows represent one of the iris species. Columns of the top 3 rows show how the 3 classes were distributed among the clusters, where each column represents a cluster. The bottom rows should be all zeros.
- Now use the predict() method and the clusters for k=3 to classify the entire iris data set.
- Print out the confusion matrix and accuracy.
  - Once again, you have to keep track of which cluster each class went to when you calculate the Accuracy Score. The easiest way to do that is to see what the majority class is in each cluster.
- Question 1: According to your results (i.e., elbow\_k), are there 3 species of iris represented in the iris data set?

## Part 2: Gaussian Mixture Models (GMM)

- Run the GMM algorithm for k = 1 through 20 and plot the AIC vs. k, where k is the number of components (n\_components). Use the aic() method to obtain the AIC. Remember to use "diag" as the covariance\_type parameter, not the default or your AIC curve won't look right.
- Find the "elbow" of the curve. We will call that the aic elbow k.
- Now run the GMM algorithm for k = 1 through 20 and plot the BIC vs. k, where k is the number of components (n\_components). Use the bic() method to obtain the BIC.
- Find the "elbow" of the curve. We will call that the bic\_elbow\_k.
- Now use the predict() method and the components for k=aic\_elbow\_k to classify the entire iris data set.
- Print out the confusion matrix and accuracy.
  - When you examine the knee of the curve for AIC, you may find an aic\_elbow\_k different than 3. That is perfectly fine. There is no right way to determine k.
  - o If you selected aic\_elbow\_k=3, then you have to keep track of which cluster each class went to when you calculate the Accuracy Score. The easiest way to do that is to see what the majority class is in each cluster.
  - o If you selected aic\_elbow\_k not equal to 3, then you cannot calculate an Accuracy Score. There are similar measures, but we didn't go over them in class, so I don't expect you to calculate them. Instead, print out a message something like this: "Cannot calculate Accuracy Score because the number of classes is not the same as the number of clusters".

- O You CAN, however, print out the Confusion Matrix, if you selected aic\_elbow\_k!=3 using the built-in Confusion Matrix function. The top 3 rows represent one of the iris species. Columns of the top 3 rows show how the 3 classes were distributed among the clusters, where each column represents a cluster. The bottom rows should be all zeros.
- Now use the predict() method and the components for k=bic\_elbow\_k to classify the entire iris data set.
- Print out the confusion matrix and accuracy.
  - When you examine the knee of the curve for BIC, you may find a bic\_elbow\_k different than 3. That is perfectly fine. There is no right way to determine k.
  - o If you selected bic\_elbow\_k=3, then you have to keep track of which cluster each class went to when you calculate the Accuracy Score. The easiest way to do that is to see what the majority class is in each cluster.
  - o If you selected bic\_elbow\_k not equal to 3, then you cannot calculate an Accuracy Score. There are similar measures, but we didn't go over them in class, so I don't expect you to calculate them. Instead, print out a message something like this: "Cannot calculate Accuracy Score because the number of classes is not the same as the number of clusters".
  - O You CAN, however, print out the Confusion Matrix, if you selected bic\_elbow\_k!=3 using the built-in Confusion Matrix function. The top 3 rows represent one of the iris species. Columns of the top 3 rows show how the 3 classes were distributed among the clusters, where each column represents a cluster. The bottom rows should be all zeros.
- Now use the predict() method and the components for k=3 to classify the entire iris data set.
- Print out the confusion matrix and accuracy.
  - Once again, you have to keep track of which cluster each class went to when you calculate the Accuracy Score. The easiest way to do that is to see what the majority class is in each cluster.
- Question 2a: According to your AIC results (i.e., aic\_elbow\_k), are there 3 species of iris represented in the iris data set?
- Question 2b: According to your BIC results (i.e., bic\_elbow\_k), are there 3 species of iris represented in the iris data set?

## Remember:

- Your Programming Assignments are individual-effort.
- You can brainstorm with other students and help them work through problems in their programs, but everyone should have their own unique assignment programs.