## **Machine Learning**

# KNNC Olivetti Faces Data – Dimensionality Reduction using Principal Components

This is a classification implementation using KNNC.

(a) Download the Olivetti faces dataset. There are 40 classes (corresponding to 40 people), each class having 10 faces of the individual; so there are a total of 400 images. Here each face is viewed as an image of size  $64 \times 64$  (= 4096) pixels; each pixel having values 0 to 255 which are ultimately converted into floating numbers in the range [0,1]. Visit https://scikit-learn.org/0.19/datasets/olivetti faces.html for more details.

Task 2: Here, you need to use **bootstrapping** to generate 10 more training patterns from each class (person), as follows:

- (a) Let  $\mathcal{X}$  be the training dataset of 400 face images.
- (b) Let the set *RESAMPLES* be empty.
- (c) For each of the training patterns  $X_i \in \mathcal{X}$  (for i = 1, ..., 400) do the following:
  - i. Let  $X_i$  be the training pattern.
  - ii. Let  $X_i^1, X_i^2, \dots, X_i^P$  be the P nearest neighbors of  $X_i$  from the **remaining** patterns of the same class as that of  $X_i$ .
  - iii. Let

$$X_i' = \frac{1}{P+1} \sum_{j=0}^{P} X_i^j,$$

where  $X_i^0 = X_i$  itself.

iv. Add  $X'_i$  to set RESAMPLES.

(d) Note that there are 400 patterns in  $\mathcal{X}$ . Obtain 400 more in RESAMPLES using P=3. Now update  $\mathcal{X}$  as

$$\mathcal{X} = \mathcal{X} \cup RESAMPLES.$$

**Task 5:** In this task you are supposed to reduce the dimensionality using I principal components for the values of I = 200; 400; 600; 800 on the dataset obtained in task 2 step(d). Compute the *KNNC* accuracy for different values of K and different distances as in task 4.

#### CODE:

Please find the code committed for PCA as

KNNC OlivettiFaceData Task5 Dimentionality Reduction PCA impl.py

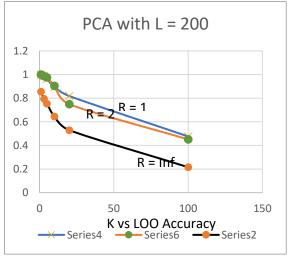
- X resampled data is first obtained using the bootstrapping method mentioned in TASK2.
- **PCA** is used to reduce the dimensionality of the data from 4096 to given values of l=200,400,600 and 800.
- With reduced dimensionality, KNNC is applied for different values of K and R and LOO accuracy is computed.

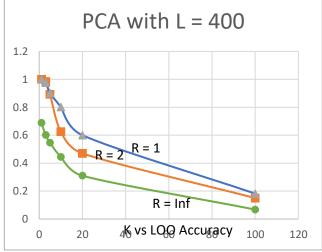
#### **RESULT:**

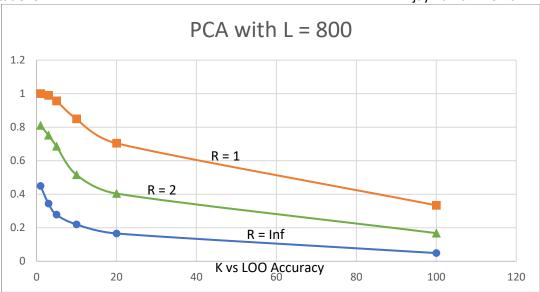
- The result is tabulated for all the L PCA values listed in the problem.
- Accuracy and LOO accuracy values are also tabulated and corresponding plots are generated.

X≡ Output\_file\_Task5.xls Result sheet is attached for all values of L. Example for L = 200 Leave on out r - (exp **Accuracy** value in (Resamples, Minkowski n=800 distance) samples) K **Accuracy** 1 0.991667 1 3 0.954167 0.986 5 0.963 0.941667 1 10 0.808333 0.894 20 0.818 0.666667 100 0.333333 0.476 1 0.991667 1 3 0.958333 0.986 5 0.920833 0.976 200 2 10 0.754167 0.904 20 0.5375 0.749 100 0.258333 0.451 1 8.0 0.855 3 0.708333 0.794 5 0.629167 0.754 Inf 10 0.533333 0.645 20 0.404167 0.529 100 0.1625 0.216

### **PLOTS:**







# **INFERENCE/ANALYSIS:**

- The task here required to reduce the image dimensionality from 4096 to various L values like 200,400,600 and 800 Principal components.
- PCA is implemented using sklearn and input is transformed to L values and KNNC is applied different K and R values.
- The result shows that lesser values of K (K<10) are sufficient to provide higher accuracy when PCA transformation is used on the data.
- With lesser K values, the usage will be more optimized to predict the test values.

#### **RESOURCES USED FOR THE ASSIGNMENT:**

• Environment:

Anaconda, Jupyter notebook

• Software:

Python

Python libraries/modules: Pandas, Numpy, SkLearn etc