PCA Code

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# Compute a PCA
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n_components = 100 ##number of components to retain

 $pca = PCA(n_components = n_components, whiten = True).fit(X_train) \quad \textit{##learning PCA decision boundary}$

apply PCA transformation

X_train_pca = pca.transform(X_train) ## project the training data into lower dimensional space

 $X_{\text{test_pca}} = \text{pca.transform}(X_{\text{test}}) \# \text{project the test data into lower dimensional space}$

train a neural network (using a neural network on the training data)

print("Fitting the classifier to the training set")

clf = MLPClassifier(hidden_layer_sizes=(1024,), batch_size=256, verbose=True, early_stopping=True).fit(X_train_pca, y_train)

Example in the link: https://pythonmachinelearning.pro/face-recognition-with-eigenfaces/

LDA Code

from sklearn.discriminant analysis import LinearDiscriminantAnalysis as LDA

lda = LDA(n_components=1) ## calling LDA function and stating the number of components to retain

X_train = lda.fit_transform(X_train, y_train) ## learning LDA decision boundaries and

 $X_{\text{test}} = \text{lda.transform}(X_{\text{test}})$ ## projecting the original test data into the transformed space (also called as fisher space)

Training a random forest classifier on the data reduced by LDA

from sklearn.ensemble import RandomForestClassifier

classifier = RandomForestClassifier(max_depth=2, random_state=0)

classifier.fit(X_train, y_train)

y_pred = classifier.predict(X_test)

Example in this link: https://stackabuse.com/implementing-lda-in-python-with-scikit-learn/