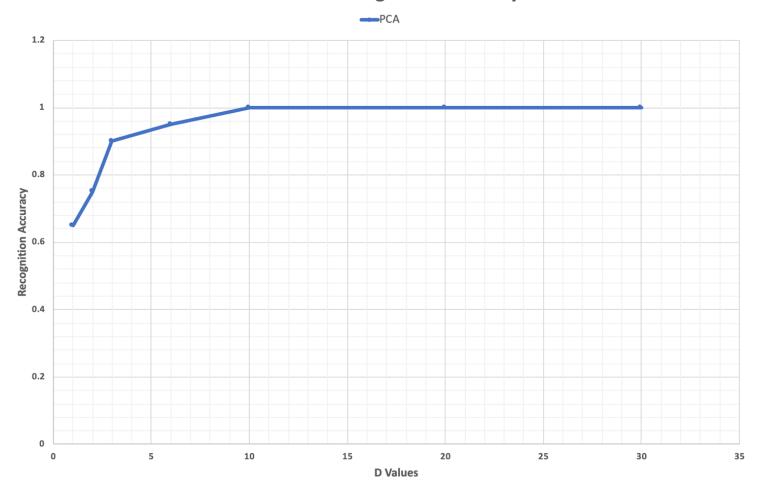
COEN 240 Machine Learning

Homework #6

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Problem 1.1

D Value Vs Recognition Accuracy

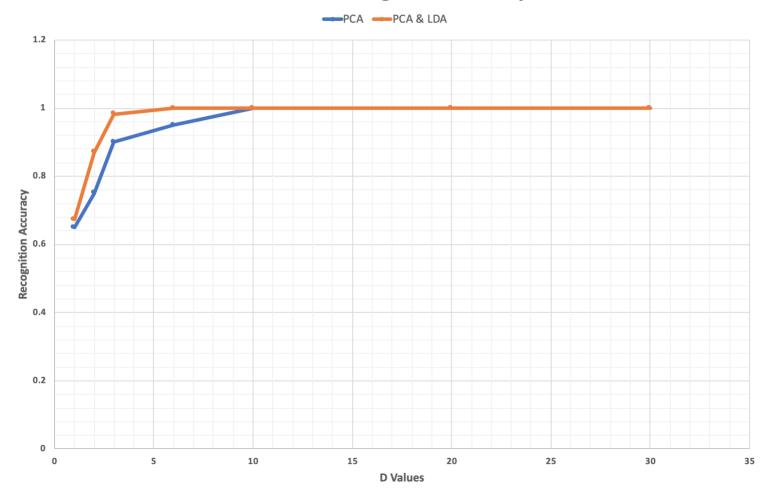


Comment: For PCA alone, trying different d values from 1, 2, 3, 6,10, 20, 30 proved to show an increase in recognition accuracy from a low of 65% to eventually getting 100% accuracy.

PCA Accuracy Results - 0.65, 0.75, 0.9, 0.95, 1.0, 1.0, 1.0

Problem 1.2

D Value Vs Recognition Accuracy



Comment: For PCA and LDA, trying different d values from 1, 2, 3, 6,10, 20, 30 proved to show an increase in recognition accuracy from a low of 67.25% to eventually getting 100% accuracy quicker than the PCA only model. Especially at lower dimensions, the accuracy is improved with the PCA and LDA model.

PCA Accuracy Results - 0.65, 0.75, 0.9, 0.95, 1.0, 1.0, 1.0 PCA & LDA Accuracy Results - 0.6725, 0.87, 0.9825, 1.0, 1.0, 1.0, 1.0

Attachment

Problem 1.1 and 1.2 Code (in zip file):

```
import os
import numpy as np
import pandas as pd
from skimage.io import imread
from sklearn.model selection import train test split
from sklearn.neighbors import KNeighborsClassifier
from sklearn.decomposition import PCA
from sklearn.discriminant analysis import LinearDiscriminantAnalysis
# get working directory plus folder of pictures
current_dir = os.getcwd() + '/att_faces_10/'
# init samples and actual y values
samples = []
y = []
# read in folder by folder samples and y values
for i in range (1, 11):
    folder path = current dir +'s' + str(i)
    image list = os.listdir(folder path)
    for im in image list:
        image = imread(folder path + '/' + im)
        samples.append(image.flatten())
        y.append(i)
# init lists and variables for d and d0
d0 = 40
d \text{ vals} = [1, 2, 3, 6, 10, 20, 30]
recognition accuracy rates pca = []
recognition accuracy rates pca lda = []
# init lda constant pca variable
pca0 = PCA(n components=d0)
for d in d vals:
    # init variables
    total calculations = 0
    correct classifications_pca = 0
    correct classifications pca lda = 0
    # init lda models
```

```
lda = LinearDiscriminantAnalysis(n components=d) # FLD /LDA
    for i in range (0, 20):
        # initializing knn models
       knnModel pca = KNeighborsClassifier(n neighbors=1,
metric='euclidean')
       knnModel pca lda = KNeighborsClassifier(n neighbors=1,
metric='euclidean')
        # splitting each class into train and test sets
       X train1, X test1, y train1, y test1 =
train test split(samples[0:10], y[0:10], test size=0.2, random state=25)
        X train2, X test2, y train2, y test2 =
train test split(samples[10:20], y[10:20], test size=0.2, random state=25)
       X train3, X test3, y train3, y test3 =
train_test_split(samples[20:30], y[20:30], test_size=0.2, random_state=25)
       X_train4, X_test4, y_train4, y_test4 =
train test split(samples[30:40], y[30:40], test_size=0.2, random_state=25)
        X_train5, X_test5, y_train5, y_test5 =
train test split(samples[40:50], y[40:50], test size=0.2, random state=25)
       X train6, X test6, y train6, y test6 =
train test split(samples[50:60], y[50:60], test size=0.2, random state=25)
       X train7, X test7, y train7, y test7 =
train test split(samples[60:70], y[60:70], test_size=0.2, random_state=25)
        X_train8, X_test8, y_train8, y_test8 =
train test split(samples[70:80], y[70:80], test size=0.2, random state=25)
        X train9, X test9, y train9, y test9 =
train test split(samples[80:90], y[80:90], test size=0.2, random state=25)
        X train10, X test10, y train10, y test10 =
train test split(samples[90:100], y[90:100], test size=0.2, random state=25)
        # combining all training and test sets
        x train = X train1 + X train2 + X train3 + X train4 + X train5 +
X train6 + X train7 + X train8 + X train9 + X train10
        y_train = y_train1 + y_train2 + y_train3 + y_train4 + y_train5 +
y train6 + y train7 + y train8 + y train9 + y train10
       x_{test} = X_{test1} + X_{test2} + X_{test3} + X_{test4} + X_{test5} + X_{test6} +
X_test7 + X_test8 + X_test9 + X_test10
       y test = y test1 + y test2 + y test3 + y test4 + y test5 + y test6 +
y test7 + y test8 + y test9 + y test10
        # running pca analysis using specific d value
       pca = PCA(n_components=d)
       pca_operator = pca.fit(x_train)
       L0 pca = pca operator.transform(x train)
```

```
# creating knn model and predicting values
        knnModel pca.fit(L0 pca, y train)
        y pred pca = knnModel pca.predict(pca operator.transform(x test))
        # creating pca using d0=40 for input to lda
        pca0 operator = pca0.fit(x train)
        L0 = pca0 operator.transform(x train)
        # create lda operation from pca
        lda operator = lda.fit(L0, y train)
        train proj lda = lda operator.transform(L0) # columns are examples
        # predict using knn
        knnModel pca lda.fit(train proj lda, y train)
        y pred pca lda =
knnModel pca lda.predict(lda operator.transform(pca0 operator.transform(x tes
t)))
        # counting predictions and total classifications
        for i in range(0, len(y pred pca lda)):
            if y pred pca lda[i] == y test[i]:
                correct classifications pca lda +=1
            if y pred pca[i] == y test[i]:
                correct_classifications_pca +=1
            total calculations +=1
    # append final results to list
recognition accuracy rates pca.append(correct classifications pca/total calcu
lations)
recognition accuracy rates pca lda.append(correct classifications pca lda/tot
al calculations)
# print out rates
print(recognition_accuracy_rates_pca)
print(recognition accuracy rates pca lda)
```