IDC-409 Project-2 Signature Fraud Busting

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1 Contribution

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2 Problem

You have a collection of signatures. Given a new signature, you are asked to find out if it is a real signature or a fake one. Write a possible solution using K-means or any other clustering method.

3 Source of dataset

The data has been taken from Kaggle.

The dataset available on Kaggle for signature verification includes a collection of authentic and forged signatures of a number of people which can be used for training and testing the algorithm.

4 Data pre-processing

We have used tensorflow and keras for pre-processing the images.

Inception v3 assigns weights pre-trained in ImageNet data for extracting the image features. The ImageNet data set is a collection of images belonging different categories and labels on which the module has been trained.

After extracting the features of each image, we extracted arrays from each image. The arrays were then pre-processed. After training, we used predict() to predict the labels and the data was flattened to 1 dimension.

We created a pandas data frame to show the name of each image with its corresponding cluster.

5 Approach to the problem

The algorithm used is k-means.

Since in the given problem, we had to use clustering methods and not the supervised learning methods, we took the collection of signatures of a single person. The data set contained both - the authentic signatures and the forged ones. We were aware which signatures were authentic as given in the problem.

On the basis of this information, we try to apply k means on the data with k=2. Since we already knew which data points are the authentic signatures, by looking at the clusters we could identify the forged signatures.

For testing the algorithm, we first supplied only the authentic data and observed that the algorithm clustered all the points in a single cluster indicating there was no significant difference between them.

After the testing, we ran the algorithm on a collection of authentic and forged data sets. The algorithm clustered the two separately. Since we were already aware which were the authentic ones, the problem of forged signature busting was solved.

6 Results

6.1 The code for the problem is as follows -

```
#Importing the required libraries
2
   from sklearn.cluster import KMeans
3
   import os
   import pandas as pd
   import numpy as np
   from tqdm import tqdm
   from \ tensorflow.keras.applications.inception\_v3 \ import \ Inception V3
   from tensorflow.keras.preprocessing import image
10
11
   from tensorflow.keras.preprocessing.image import img_to_array
   from tensorflow.keras.applications.inception_v3 import preprocess_input
12
13
   #Function for pre-processing and extracting the image features
14
15
   def image_features(directory):
16
       train = InceptionV3(weights='imagenet', include_top=False)
17
       #for pre-processing, pre defined weights used for identifying the
18
       object from the image
       features = [] # list of features extracted
19
       image_name = [] # the image names
20
        for i in tqdm(directory):
21
            file_name=' '+i # file name format in which images are present in
22
        the directory
            img=image.load_img(file_name, target_size=(224,224))
23
            x = img\_to\_array(img) #converting the image to array
24
            x=np.expand_dims(x,axis=0) # expanding the array for adding cluster
25
        information
            x=preprocess_input(x) # pre-processing the arrays
26
            feature=train.predict(x) # testing
27
            feature = feature \,.\, flatten \,(\,) \quad \# \ dimension \ reduction
28
            features.append(feature) # adding the features to the list
29
            image_name.append(i) #adding the image name to the list
        return features, image_name
```

```
#Getting the images from the directory, extracting the features and making
33
       the dataframe
34
   file_path = os.getcwd() # the working directory
35
   #print(path)
   image_path=os.listdir('006') #the files are named as '006..'
37
   #print(img_path)
38
   #extracting the features using the function defined in previous cell
39
   image_features , image_name=image_feature(image_path)
40
   # creating the pandas dataframe for the images
41
   image_cluster = pd.DataFrame(image_name, columns=['Image name'])
42
43
   #print(image_cluster)
44
45
   #Applying k means
46
   k=2 #number of clusters to be formed
47
       initial state for each iteration is chosen randomly
48
       max. number of iterations allowed for a single run = 10000000
49
       tolerance in clustering = 1e-15
50
       Number of times the algorithm is run with different centroids =500
51
52
53
   clusters = KMeans(k, random_state =None, max_iter=10000000, tol=1e-15, n_init
54
   clusters fit (image_features) #computing the clusters from the extracted
55
       features
   X = clusters.fit_transform(image_features) # computing the distance from
       centroids
57
   print(X)
58
   #Visualising the results
59
60
   image_cluster["Cluster Number"] = clusters.labels_ #adding labels to each
61
       cluster
   print(image_cluster)
```

6.2 The following results were obtained:

	Ттада пата	Cluster Number
0	Image name 006 01.PNG	Cluster Number
1		1
	006_02.PNG	
2	006_03.PNG	1
3	006_04.PNG	1
4	006_05.PNG	1
5	006_06.PNG	1
6	006_07.PNG	1
7	006_08.PNG	1
8	006_09.PNG	1
9	006_10.PNG	1
10	006_11.PNG	1
11	006_12.PNG	1
12	006_13.PNG	1
13	006_14.PNG	1
14	006_15.PNG	0
15	006_16.PNG	1
16	006_17.PNG	1
17	006_18.PNG	0
18	006 19.PNG	1
19	006 20.PNG	1
20	006_21.PNG	0
21	006 22.PNG	1
22	006 23.PNG	1
23	006 24.PNG	0
24	006_25.png	0
25	006_26.png	0
26	006_27.png	0
27	006_28.png	0
28	006 29.png	0
29	006_30.png	0
25	Buld.oc_ooo	0

Figure 1: Clusters

7 Interpretation

Image number 01 to 23 were authentic signatures and image number 24 to 30 were forged.

Most of the authentic images have been in one cluster and the forged ones in the other cluster.

However, three of the authentic ones - 15, 18 and 21 have been declared fake by the algorithm. Signatures have a possibility of differing from each other even when they are done by the same person. There may be subtle differences and the algorithm might be catching those.

No fake signature has been declared authentic by the algorithm.

Figure 2: Authentic signature

Figure 3: Fake signature identified as authentic

Figure 4: Fake signature identified as authentic

Johnson

Figure 5: Fake signature identified as authentic

Jolian Pas

Figure 6: Fake signature identified as fake

8 List of files attached

- 1. Code
- 2. The images used for clustering