Assignment -3

Python Programming

Assignment Date	9 october 2022
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Maximum Marks	2 Marks

Question-1:

Download the Dataset

Solution:

from google.colab import drivedrive.mount('/content/drive')



Download the Dataset

In [2]: from google.colab import drive
 drive.mount('/content/drive')

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

Question-2:

Image Augmentation

Solution:

Image Augmentation

```
In [3]: import numpy as np
          import pandas as pd
          import matplotlib.pyplot as plt
          from matplotlib import style
          import seaborn as sns
          import cv2
          import matplotlib.pyplot as plt
          import numpy as np
          import pandas as pd
          import os
          import PIL
          import random
          import cv2
          from tensorflow.keras import layers, models
          import tensorflow as tf
          import pandas as pd
          from sklearn.model_selection import train_test_split
          import seaborn as sns
          import pickle
          import zipfile
          tf.__version__
         12.8.21
 Out[3]:
 In [4]: !ls
         drive sample_data
In [5]: try:
            tpu = tf.distribute.cluster_resolver.TPUClusterResolver()
            print('Device:', tpu.master())
            tf.config.experimental_connect_to_cluster(tpu)
            tf.tpu.experimental.initialize_tpu_system(tpu)
            strategy = tf.distribute.experimental.TPUStrategy(tpu)
         except:
            strategy = tf.distribute.get_strategy()
         print('Number of replicas:', strategy.num_replicas_in_sync)
        Number of replicas: 1
In [6]: AUTOTUNE = tf.data.experimental.AUTOTUNE
         batch size = 32
         IMAGE_SIZE = [128, 128]
         EPOCHS = 25
In [7]:
         image = cv2.imread(r'/content/drive/MyDrive/Flowers-Dataset/flowers/daisy/100080576_f52e8ee070_n.jpg')
In [8]:
         print(image.shape)
        (263, 320, 3)
         imgplot = plt.imshow(image)
         plt.show()
```

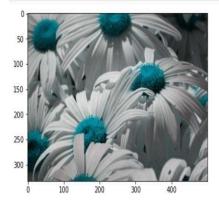
```
100 -
150 -
200 -
250 -
0 50 100 150 200 250 300
```

```
GCS PATH = "/content/drive/MyDrive/Flowers-Dataset/flowers"
                             \label{eq:class_names} $$ $ \text{CLASS\_NAMES} = np.array([str(tf.strings.split(item, os.path.sep)[-1].numpy())[2:-1] $$ $ $ \text{CLASS\_NAMES} = np.array([str(tf.strings.split(item, os.path.sep)[-1].numpy())[2:-1] $$ $ \text{CLASS\_NAMES} = np.array([str(tf.strings.split(item, os.path.sep)[-1].numpy() $$ $ \text{CLASS\_NAMES} = np.array([str(tf.strings.split(item, os.path.sep)[-1].numpy() $$ $ \text{CLASS\_NAMES} = np.array(
                                                                                              for item in tf.io.gfile.glob(str(GCS_PATH + "*/*"))])
                             CLASS_NAMES
                           array(['daisy', 'rose', 'dandelion', 'sunflower', 'tulip'], dtype='<U9')</pre>
  In [11]:
                             files_count = []
                             for i,f in enumerate(CLASS_NAMES):
                                       folder_path = os.path.join(GCS_PATH, f)
                                        for path in os.listdir(os.path.join(folder_path)):
                                                   files\_count.append(['\{\}/\{\}'.format(folder\_path,path), \ f, \ i])
                             flowers_df = pd.DataFrame(files_count, columns=['filepath', 'class_name', 'label'])
                             flowers df.head()
Out[11]:
                                                                                                                      filepath class_name label
                           0 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                                                                                                                                                0
                           1 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                                                                                                                                                0
                           2 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                                                                                                                         daisy
                                                                                                                                                                                0
                           3 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                                                                                                                         daisy
                                                                                                                                                                                0
                           4 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                                                                                                                         daisy
                                                                                                                                                                                0
In [12]: flowers_df.class_name.value_counts()
                          dandelion
                                                             1052
Out[12]:
                          tulip
                                                                 984
                                                                 784
                           rose
                           daisy
                           sunflower
                                                                733
                           Name: class_name, dtype: int64
In [13]:
                             quantidade_por_class = 500
                             flowers_df = pd.concat([flowers_df[flowers_df['class_name'] == i][:quantidade_por_class] for i in CLASS_NAMES])
In [14]:
                             flowers_df.class_name.value_counts()
                         daisy
                                                              500
Out[14]:
                           rose
                                                              500
                           dandelion
                                                              500
                           sunflower
                                                             500
                           tulip
```

sunflower 500 tulip 500

Name: class_name, dtype: int64

```
In [15]:
    image = cv2.imread(flowers_df.filepath[100])
    imgplot = plt.imshow(image)
    plt.show()
```



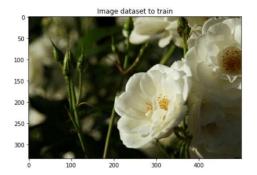
Create Model

```
x_train, x_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=101)
```

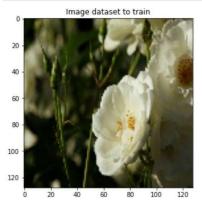
3:

Create Model

Solution:



```
100 - 120 - 50 100 150 200
```



```
100 - 20 40 60 80 100 120
```

```
In [23]:
    train_batches = train_data_norm.batch(batch_size)
    test_batches = test_data_norm.batch(batch_size)

for i, l in train_batches.take(1):
    print('Train Data Shape',i.shape)
    for i, l in test_batches.take(1):
        print('Test Data Shape',i.shape)
```

Train Data Shape (32, 128, 128, 3) Test Data Shape (32, 128, 128, 3)

4 :

Add Layers (Convolution, MaxPooling, Flatten, Dense-(Hidden Layers), Output)

Solution:

Add Layers (Convolution, MaxPooling, Flatten, Dense-(Hidden Layers), Output)

```
LeNet = models.Sequential()
LeNet.add(layers.Conv2D(6, (5,5), activation = 'relu', input_shape = (128, 128, 3)))
LeNet.add(layers.MaxPooling2D())
LeNet.add(layers.MaxPooling2D())
LeNet.add(layers.MaxPooling2D())
LeNet.add(layers.Flatten())
LeNet.add(layers.Dense(255, activation='relu'))
LeNet.add(layers.Dense(255, activation='relu'))
LeNet.add(layers.Dense(124, activation='relu'))
LeNet.add(layers.Dense(124, activation='relu'))
LeNet.add(layers.Dense(84, activation='relu'))
LeNet.add(layers.Dense(43, activation='relu'))
LeNet.add(layers.Dense(43, activation='relu'))
LeNet.summary()
```

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 124, 124, 6)	
<pre>max_pooling2d (MaxPooling2D)</pre>	(None, 62, 62, 6)	0
conv2d_1 (Conv2D)	(None, 58, 58, 16)	2416
max_pooling2d_1 (MaxPooling 2D)	(None, 29, 29, 16)	0
flatten (Flatten)	(None, 13456)	0
dense (Dense)	(None, 255)	3431535
dropout (Dropout)	(None, 255)	0
dense_1 (Dense)	(None, 124)	31744
dropout_1 (Dropout)	(None, 124)	0
dense_2 (Dense)	(None, 84)	10500
dense_3 (Dense)	(None, 43)	3655

Question-5:

Non-trainable params: 0

Compile The Model

Solution:

Compile The Model

6: Fit

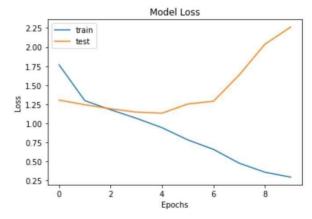
The Model

Solution:

Question-

Fit The Model

```
In [26]: history = LeNet.fit(train_batches, epochs=10,batch_size = 16,validation_data=(test_batches))
     Epoch 1/10
     55/55 [====
Epoch 2/10
             55/55 [====
             Epoch 3/10
                  ========] - 42s 752ms/step - loss: 1.1785 - accuracy: 0.5034 - val_loss: 1.1907 - val_accuracy: 0.5173
     Epoch 4/10
     Epoch 4/10
55/55 [=====
Epoch 5/10
55/55 [=====
Epoch 6/10
            :==========] - 49s 889ms/step - loss: 0.9430 - accuracy: 0.6366 - val_loss: 1.1333 - val_accuracy: 0.5520
     55/55 [====
Epoch 7/10
55/55 [====
               =============================== ] - 36s 648ms/step - loss: 0.6586 - accuracy: 0.7531 - val_loss: 1.2900 - val_accuracy: 0.5427
     Epoch 8/10
     55/55 [====
                  =========] - 40s 719ms/step - loss: 0.4778 - accuracy: 0.8257 - val_loss: 1.6341 - val_accuracy: 0.5080
     Epoch 9/10
55/55 [====
                Epoch 10/10
     In [31]:
        plt.plot(history.history['loss'])
        plt.plot(history.history['val_loss'])
        plt.title('Model Loss')
        plt.ylabel('Loss')
        plt.xlabel('Epochs')
        plt.legend(['train', 'test'])
        plt.show()
```



Question-7: Save the Model

Solution:

Save the Model

```
In [32]:
          from sklearn.neighbors import KNeighborsClassifier as KNN
          import numpy as np
          # Load dataset
          from sklearn.datasets import load_iris
          iris = load_iris()
          X = iris.data
          y = iris.target
          # Split dataset into train and test
          X_train, X_test, y_train, y_test = \
              train_test_split(X, y, test_size=0.3,
                               random_state=2018)
          # import KNeighborsClassifier model
          knn = KNN(n_neighbors=3)
          # train model
          knn.fit(X_train, y_train)
         KNeighborsClassifier(n_neighbors=3)
Out[32]:
In [30]:
          import pickle
          saved_model = pickle.dumps(knn)
          knn_from_pickle = pickle.loads(saved_model)
          knn_from_pickle.predict(X_test)
         array([0, 1, 1, 1, 0, 1, 2, 1, 2, 0, 0, 2, 2, 2, 0, 2, 2, 0, 1, 1, 1, 0,
Out[30]:
                2, 0, 0, 2, 0, 0, 2, 1, 0, 2, 0, 1, 2, 0, 0, 0, 0, 0, 1, 0, 2, 2, 2,
```

Question-8:

Test The Model

Solution:

Test The Model

```
In [27]:
           import warnings
           warnings.filterwarnings('always')
           warnings.filterwarnings('ignore')
In [28]:
           plt.plot(history.history['accuracy'])
           plt.plot(history.history['val_accuracy'])
           plt.title('Model Accuracy')
           plt.ylabel('Accuracy')
           plt.xlabel('Epochs')
           plt.legend(['train', 'test'])
           plt.show()
                              Model Accuracy
      0.9
                                                                                               0.9
                train
                test
      0.8
                                                                                               0.8
                                                                                               0.7
      0.7
   Accuracy
                                                                                            Accuracy
      0.6
                                                                                               0.6
      0.5
                                                                                               0.5
                                                                                               0.4
      0.4
      0.3
                                                                                               0.3
                       ż
            0
                                              6
                                                         8
                                   Epochs
```