Assignment -3

Python Programming

Assignment Date	13 OCTOBER 2022
Student Name	SANTHOSH K
Student Roll Number	111519104126
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__
Out[3]: '2.8.2'
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
         AUTOTUNE = tf.data.experimental.AUTOTUNE
         batch_size = 32
         IMAGE_SIZE = [128, 128]
         EPOCHS = 25
         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 -

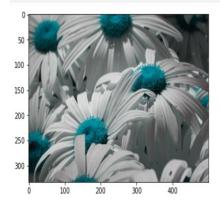
0 50 100 150 200 250 300
```

```
In [10]:
            GCS_PATH = "/content/drive/MyDrive/Flowers-Dataset/flowers"
            \label{eq:class_names} {\tt CLASS\_NAMES} = {\tt np.array}([{\tt str(tf.strings.split(item, os.path.sep})[-1].numpy())[2:-1]
                                      for item in tf.io.gfile.glob(str(GCS_PATH + "*/*"))])
            CLASS_NAMES
 {\tt Out[10]:} \  \  \, {\tt array(['daisy', 'rose', 'dandelion', 'sunflower', 'tulip'], \ dtype='<U9')}
 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...
          1 /content/drive/MyDrive/Flowers-Dataset/flowers...
          2 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                                       0
          3 /content/drive/MyDrive/Flowers-Dataset/flowers...
          4 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                                       0
                                                              daisy
In [12]:
           flowers_df.class_name.value_counts()
          dandelion
                        1052
Out[12]:
          tulip
                          984
                          784
                          764
          daisy
                         733
          sunflower
          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()
                         500
          daisy
Out[14]:
                         500
          dandelion
                         500
          sunflower
                         500
          tulip
                         500
```

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

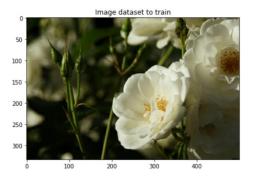
```
In [16]: X = flowers_df['filepath']
y = flowers_df['label']

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

Question-3:

Create Model

Solution:

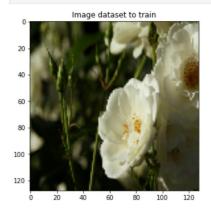


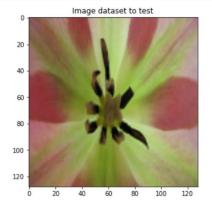
```
100 - 100 - 100 - 150 200
```

```
In [21]:
    def preprocessing(image, label):
        returns a image that is reshaped and normalized
        """
        image = tf.cast(image, tf.float32)
        image = image / 255.
        image = tf.image.resize(image, IMAGE_SIZE)
        return image, label

        train_data_norm = train_data_img.map(preprocessing)
        test_data_norm = test_data_img.map(preprocessing)

In [22]:
        fig, ax = plt.subplots(1,2, figsize = (15,5))
        for i,l in train_data_norm.take(1):
            ax[0].set_title('Image dataset to train');
            ax[0].imshow(i);
        for i,l in test_data_norm.take(1):
            ax[1].set_title('Image dataset to test');
        ax[1].imshow(i);
```





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

    for i, 1 in train_batches.take(1):
        print('Train_Data_Shape',i.shape)
    for i, 1 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)
```

Question-4:

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

Solution:

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

```
In [24]:
    LeNet = models.Sequential()
    LeNet.add(layers.Conv2D(6, (5,5), activation = 'relu', input_shape = (128, 128, 3)))
    LeNet.add(layers.MaxPooling2D())
    LeNet.add(layers.Conv2D(16, (5,5), activation = 'relu'))
    LeNet.add(layers.MaxPooling2D())
    LeNet.add(layers.Flatten())
    LeNet.add(layers.Dense(255, activation='relu'))
    LeNet.add(layers.Dropout(0.2))
    LeNet.add(layers.Dense(124, activation='relu'))
    LeNet.add(layers.Dropout(0.2))
    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)		
<pre>max_pooling2d (MaxPooling2D)</pre>	(None, 62, 62, 6)	0
conv2d_1 (Conv2D)	(None, 58, 58, 16)	2416
<pre>max_pooling2d_1 (MaxPooling 2D)</pre>	(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
Total params: 3,480,306 Trainable params: 3,480,306 Non-trainable params: 0		

Question-5:

Compile The Model

Solution:

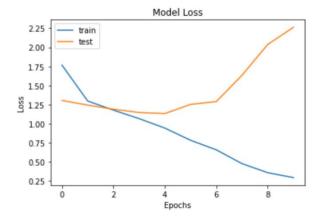
Compile The Model

Question-6: Fit The Model

Solution:

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
                     ================] - 130s 2s/step - loss: 1.7673 - accuracy: 0.2943 - val_loss: 1.3046 - val_accuracy: 0.4560
       55/55 [====
Epoch 3/10
                      ==========] - 40s 724ms/step - loss: 1.2971 - accuracy: 0.4434 - val_loss: 1.2441 - val_accuracy: 0.4880
       55/55 [====
                       ==========] - 42s 752ms/step - loss: 1.1785 - accuracy: 0.5034 - val_loss: 1.1907 - val_accuracy: 0.5173
       Epoch 4/10
       55/55 [====
Epoch 5/10
                       55/55 [======
                     Froch 6/10
       55/55 [====
Epoch 7/10
                         ========] - 37s 673ms/step - loss: 0.7835 - accuracy: 0.7051 - val_loss: 1.2531 - val_accuracy: 0.5333
                         ========] - 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 [=====
                    ============================ ] - 36s 647ms/step - loss: 0.3595 - accuracy: 0.8703 - val_loss: 2.0376 - val_accuracy: 0.4947
       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)
Out[30]: array([0, 1, 1, 1, 0, 1, 2, 1, 2, 0, 0, 2, 2, 2, 0, 2, 2, 0, 1, 1, 1, 0,
                2, 0, 0, 2, 0, 0, 2, 1, 0, 2, 0, 1, 2, 0, 0, 0, 0, 1, 0, 2, 2, 2,
                1])
```

Question-8: Test The Model

Solution:

Test The Model

```
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()
```

