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__
 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)
             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)
In [9]:
         imgplot = plt.imshow(image)
         plt.show()
```

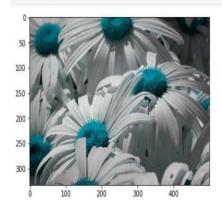
```
150 -
200 -
250 -
0 50 100 150 200 250 300
```

```
In [10]:
          GCS_PATH = "/content/drive/MyDrive/Flowers-Dataset/flowers"
          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...
                                                                  0
                                                          daisy
          1 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                          daisy
          2 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                                  0
                                                          daisy
          3 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                          daisy
          4 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                                  0
                                                          daisy
 In [12]:
           flowers_df.class_name.value_counts()
Out[12]: dandelion
                       1052
          tulip
                         984
                         784
          rose
          daisy
                         764
          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()
 Out[14]: daisy
                        500
                        500
           rose
           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

```
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)
```

Tn [17].

3:

Create Model

Solution:

```
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)

In [17]: x_train_tensor = tf.convert_to_tensor(x_train.values, dtype=tf.string)
y_train_tensor = tf.convert_to_tensor(y_train.values)

x_test_tensor = tf.convert_to_tensor(x_test.values, dtype=tf.string)
y_test_tensor = tf.convert_to_tensor(x_test.values)

In [18]: train_data = tf.data.Dataset.from_tensor_slices((x_train_tensor, y_train_tensor))
test_data = tf.data.Dataset.from_tensor_slices((x_test_tensor, y_test_tensor))

In [19]: def map_fn(path, label):
    image = tf.image.decode_jpeg(tf.io.read_file(path))
    return image, label

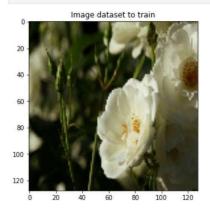
#apply the function
train_data_img = train_data.map(map_fn)
test_data_img = test_data.map(map_fn)

In [20]: fig, ax = plt.subplots(1,2, figsize = (15,5))
for i,l in train_data_img.take(1):
    ax[0].set_title('Image dataset to train');
    ax[0].imshow(1);
for i,l in test_data_img.take(1):
    ax[1].imshow(1);
```



```
Image dataset to test

20 -
40 -
60 -
80 -
100 -
120 -
0 50 100 150 200
```



```
100 - 120 - 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, 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)

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='sigmoid'))
         LeNet.summary()
                                                       Param #
                              Output Shape
   Layer (type)
  (None, 124, 124, 6)
   conv2d (Conv2D)
                                                       456
```

```
max_pooling2d (MaxPooling2D (None, 62, 62, 6)
conv2d_1 (Conv2D)
                        (None, 58, 58, 16)
                                              2416
max_pooling2d_1 (MaxPooling (None, 29, 29, 16)
                        (None, 13456)
flatten (Flatten)
dense (Dense)
                        (None, 255)
                                               3431535
dropout (Dropout)
                        (None, 255)
dense_1 (Dense)
                        (None, 124)
                                              31744
dropout_1 (Dropout)
                       (None, 124)
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

6: Fit

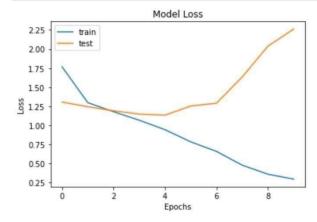
The Model

Solution:

Question-

Fit The Model

```
In [26]:
       history = LeNet.fit(train_batches, epochs=10,batch_size = 16,validation_data=(test_batches))
       55/55 [====
Epoch 2/10
                   55/55 [====
                  Epoch 3/10
55/55 [====
                        ========] - 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 [============] - 36s 650ms/step - loss: 1.0667 - accuracy: 0.5526 - val_loss: 1.1468 - val_accuracy: 0.5453
Epoch 5/10
55/55 [================] - 49s 889ms/step - loss: 0.9430 - accuracy: 0.6366 - val_loss: 1.1333 - val_accuracy: 0.5520
       Epoch 6/10
55/55 [====
Epoch 7/10
55/55 [====
                      =========] - 37s 673ms/step - loss: 0.7835 - accuracy: 0.7051 - val_loss: 1.2531 - val_accuracy: 0.5333
                 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
55/55 [====
                   :============================== ] - 41s 744ms/step - loss: 0.2947 - accuracy: 0.9023 - val_loss: 2.2624 - val_accuracy: 0.4693
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, 1, 0, 2, 2, 2,
                1])
```

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.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
                                          Epochs
```