Assignment -3

Student Roll Number	PNT2022TMID03497
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

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]: ||1s
        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)
In [9]: imgplot = plt.imshow(image)
         plt.show()
```

```
50 -

100 -

200 -

250 -

0 50 100 150 200 250 300
```

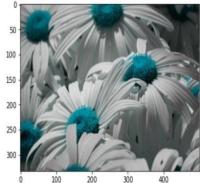
tulip

500

```
In [10]: GCS_PATH = "/content/drive/MyDrive/Flowers-Dataset/flowers"
              CLASS_NAMES = np.array([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='<09')}
  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...
                                                              daisy
                                                                       0
           1 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                              daisy
           2 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                              daisy
           3 /content/drive/MyDrive/Flowers-Dataset/flowers...
           4 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                              daisy
In [12]: flowers_df.class_name.value_counts()
Out[12]: dandelion 1052
          tulip
                          984
                          784
           rose
           daisy
                          764
                          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()
Out[14]: daisy
                         500
                         500
          rose
           dandelion
                         500
           sunflower
                         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

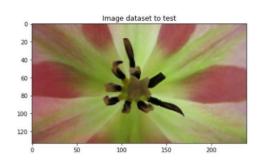
```
In [16]:
X = flowers_df['filepath']
y = flowers_df['label']

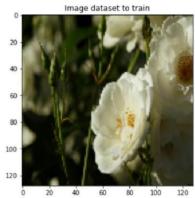
x_train, x_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=101)
```

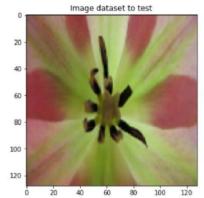
Question-3: Create

Model









```
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)

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.Dense(84, activation='relu'))
    LeNet.add(layers.Dense(84, activation='relu'))
    LeNet.add(layers.Dense(43, activation='relu'))
    LeNet.add(layers.Dense(43, activation='sigmoid'))
    LeNet.summary()
```

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 124, 124, 6)	456
<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:

Question-6:

Fit The Model

Solution:

Compile The Model

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
                            :========] - 130s 2s/step - loss: 1.7673 - accuracy: 0.2943 - val_loss: 1.3046 - val_accuracy: 0.4560
                             :=========] - 40s 724ms/step - loss: 1.2971 - accuracy: 0.4434 - val_loss: 1.2441 - val_accuracy: 0.4880
                                        ==] - 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 [======

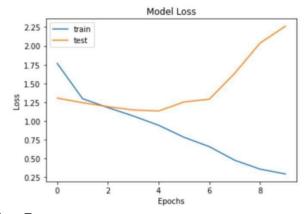
Epoch 5/10

55/55 [======

Epoch 6/10

55/55 [======

Epoch 7/10
                       ==] - 49s 889ms/step - loss: 0.9430 - accuracy: 0.6366 - val_loss: 1.1333 - val_accuracy: 0.5520
                                           - 37s 673ms/step - loss: 0.7835 - accuracy: 0.7051 - val_loss: 1.2531 - val_accuracy: 0.5333
        55/55 [===:
Epoch 8/10
                             =======] - 36s 648ms/step - loss: 0.6586 - accuracy: 0.7531 - val_loss: 1.2900 - val_accuracy: 0.5427
        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
                            ========] - 36s 647ms/step - loss: 0.3595 - accuracy: 0.8703 - val_loss: 2.0376 - val_accuracy: 0.4947
        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

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

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()
```

