# Assignment -1

# **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]: ||1s
        drive sample_data
In [5]:
            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()
          0
```

```
100 -

150 -

200 -

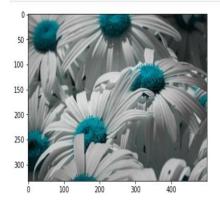
250 -

0 50 100 150 200 250 300
```

```
In [10]:
             GCS_PATH = "/content/drive/MyDrive/Flowers-Dataset/flowers"
             \label{eq:class_names} {\tt CLASS\_NAMES = 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
 Out[10]: 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
           {\bf 0} \quad / content/drive/MyDrive/Flowers-Dataset/flowers...
                                                                         0
           1 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                               daisy
           2 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                               daisy
                                                                         0
                                                               daisy
           3 /content/drive/MyDrive/Flowers-Dataset/flowers...
           4 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                               daisy
In [12]:
           flowers_df.class_name.value_counts()
          dandelion
                         1052
Out[12]:
           tulip
                          984
                          784
           rose
                          764
           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
                         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)
```

Tn [17]

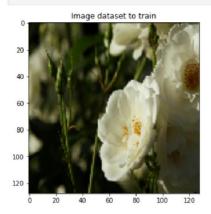
## Question-3:

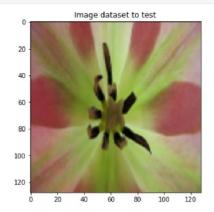
## **Create Model**

# Solution:



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





```
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)

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

Layer (type)	Output Shape	Param #
Chick Mark Special	(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:

#### Compile The Model

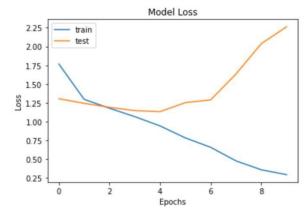
# Question-6:

#### **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 [===:
                                          =] - 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
                                ========] - 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 [====
Epoch 9/10
                               :========] - 40s 719ms/step - loss: 0.4778 - accuracy: 0.8257 - val_loss: 1.6341 - val_accuracy: 0.5080
        55/55 [====
Epoch 10/10
                               =========== - 36s 647ms/step - loss: 0.3595 - accuracy: 0.8703 - val_loss: 2.0376 - val_accuracy: 0.4947
                          ==========] - 41s 744ms/step - loss: 0.2947 - accuracy: 0.9023 - val_loss: 2.2624 - val_accuracy: 0.4693
        55/55 [=====
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, 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()
```

