## 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**

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**

```
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
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()
```

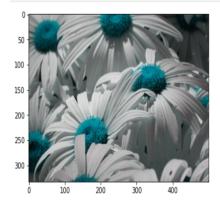
```
150 -
200 -
250 -
50 100 150 200 250 300
```

```
In [10]:
            GCS_PATH = "/content/drive/MyDrive/Flowers-Dataset/flowers"
            \label{eq: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
          0 /content/drive/MyDrive/Flowers-Dataset/flowers...
          1 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                                      0
          2 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                                      0
          3 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                                      0
                                                              daisy
          4 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                                      0
                                                              daisy
In [12]: flowers_df.class_name.value_counts()
          dandelion
                        1052
Out[12]:
          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
          rose
                         500
          dandelion
          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()
```



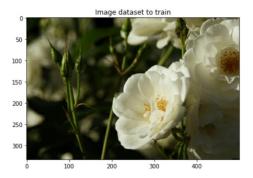
#### **Create Model**

```
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**:



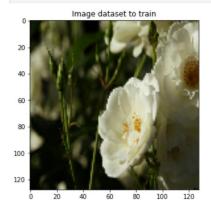
```
0 - Image dataset to test

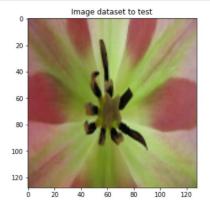
20 - 40 - 60 - 80 - 100 - 150 - 200
```

```
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.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)		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**

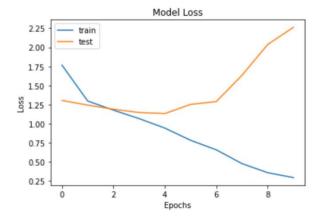
### Question-6:

### **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
                  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
                    ==========] - 36s 650ms/step - loss: 1.0667 - accuracy: 0.5526 - val_loss: 1.1468 - val_accuracy: 0.5453
      55/55 [======
                  Fnoch 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 [=====
                 Fnoch 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)
         \mathsf{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**

```
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()
```

