

**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 drive  
drive.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

```
In [6]: AUTOTUNE = tf.data.experimental.AUTOTUNE
batch_size = 32
IMAGE_SIZE = [128, 128]
EPOCHS = 25
```

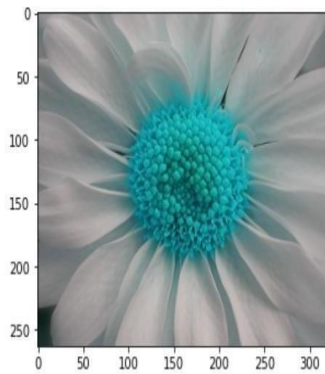
```
In [7]: image = cv2.imread(r'/content/drive/MyDrive/Flowers-Dataset/flowers/daisy/100080576_f52e8ee070_n.jpg')
```

```
In [8]: print(image.shape)
```

(263, 320, 3)

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





```
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
```

```
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...	daisy	0
1	/content/drive/MyDrive/Flowers-Dataset/flowers...	daisy	0
2	/content/drive/MyDrive/Flowers-Dataset/flowers...	daisy	0
3	/content/drive/MyDrive/Flowers-Dataset/flowers...	daisy	0
4	/content/drive/MyDrive/Flowers-Dataset/flowers...	daisy	0

```
In [12]: flowers_df.class_name.value_counts()
```

```
Out[12]: dandelion    1052
tulip          984
rose           784
daisy          764
sunflower      733
Name: class_name, dtype: int64
```

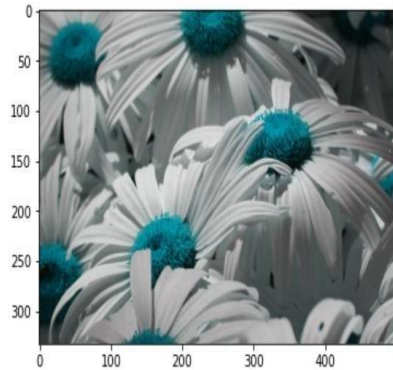
```
In [13]: quantidade_por_class = 500
flowers_df = pd.concat([flowers_df[flowers_df['class_name']== i][:quantidade_por_class] for i in CLASS_NAMES])
```

```
In [14]: flowers_df.class_name.value_counts()
```

```
Out[14]: daisy          500
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)
```

```
In [17]:
```

## Question-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(y_test.values)
```

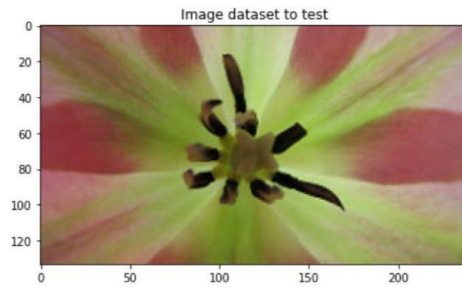
```
In [18]: train_data = tf.data.Dataset.from_tensor_slices((x_train_tensor, y_train_tensor))
test_data = tf.data.Dataset.from_tensor_slices((x_test_tensor, y_test_tensor))
```

```
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(i);
for i,l in test_data_img.take(1):
    ax[1].set_title('Image dataset to test');
    ax[1].imshow(i);
```

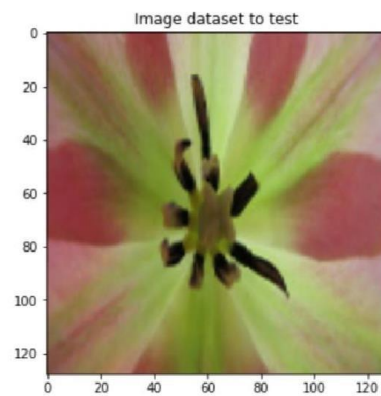


```
In [21]: def preprocessing(image, label):
          """
          returns a image that is reshaped and normalized
          """
          image = tf.cast(image, tf.float32)
          image = image / 255.
          image = tf.image.resize(image, IMAGE_SIZE)

          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, 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(43, activation='sigmoid'))
LeNet.summary()
```

Layer (type)	Output Shape	Param #
=====		
conv2d (Conv2D)	(None, 124, 124, 6)	456
max_pooling2d (MaxPooling2D)	(None, 62, 62, 6)	0
conv2d_1 (Conv2D)	(None, 58, 58, 16)	2416
max_pooling2d_1 (MaxPooling2D)	(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

```
In [25]: LeNet.compile(optimizer='Adam', loss='sparse_categorical_crossentropy',
metrics = ['accuracy'])
```

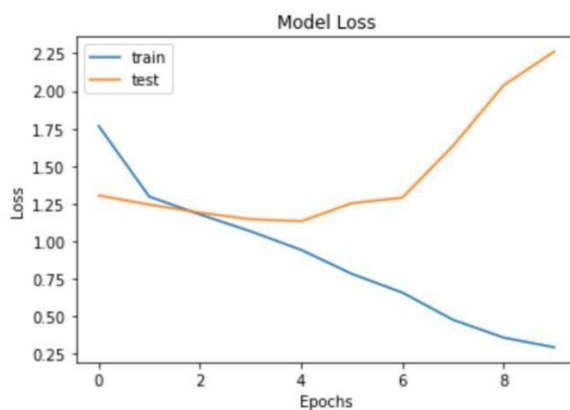


### Fit The Model

```
In [26]: history = LeNet.fit(train_batches, epochs=10, batch_size = 16, validation_data=(test_batches))
```

```
Epoch 1/10
55/55 [=====] - 130s 2s/step - loss: 1.7673 - accuracy: 0.2943 - val_loss: 1.3046 - val_accuracy: 0.4560
Epoch 2/10
55/55 [=====] - 40s 724ms/step - loss: 1.2971 - accuracy: 0.4434 - val_loss: 1.2441 - val_accuracy: 0.4880
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
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 [=====] - 37s 673ms/step - loss: 0.7835 - accuracy: 0.7051 - val_loss: 1.2531 - val_accuracy: 0.5333
Epoch 7/10
55/55 [=====] - 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
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)
```

```
Out[32]: KNeighborsClassifier(n_neighbors=3)
```

```
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

```
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()
```

