Assignment -1

Python Programming

| Assignment Date | 9 october 2022 |
|---------------------|--------------------------|
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| Student Roll Number | 111519104120 |
| Maximum Marks | 2 Marks |

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Download the Dataset

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|-----|--------|---|

from google.colab

import drivedrive.mount('/content/drive')

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| # | | | | | | | | | | | | | | | | | | | | | | | | | | 1 |

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

```
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]:
            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
```

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

sunflower

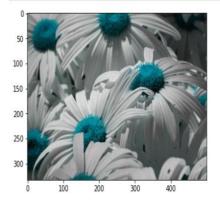
tulip

586 586

```
GCS\_PATH = "/\langle ontent/drive/NyDrive/Flowers-Dataset/flowers^{TM}
  \texttt{\pounds LASS} \ fIAhfi8 = up.array([str(th.strings.split(item, \ os.path. \ sep) \ [-1].nnpy(\ )) \ [2: \ -1]
                              for ites In tf. io.gfile.gloh(str (G£ S_PéIfl + "*/""))])
  £LASS fIASfiS
 array(['daisy', 'rose', 'dandelion', 'sunf Io|her', 'tulip'], dtype='dJ9')
  files count = []
  for i,f- 1n enumerate(CLASS_NAf 1ES) :
       fo1der_path= as.path.3oin(GCS PATA, f)
       for path In os .Jistdir(as. path. join(fol der_pat #)) :
  \label{files} files \ count.append([ ' ( )/{}' .format(fo1der_path, path), f, i]) \\ f1ouers_df = pd,Datafrane(fi1es_count, columns=[ 'fi lepath', 'c1ass_name', 'label']) \\
  Stokers df.headf \
                                        filepath class_name label
0 /c0ntent/drive/MyDrive/Flowers-Dataset/flowers...
1 /content/drive/MyDrive/flowers-Dataset/flowers...
                                                       daisy
                                                                 0
2 /content/drive/MyDrive/Flowers-Dataset/flowers...
3 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                       daisy
4 /c0ntent/drive/MyDrive/Flowers-Dataset/flowers...
                                                       daisy
                                                                 0
flouers_df.class nane.value counts()
dandelion
               1852
tulip
                 784
rose
daisy
                 764
sunflowr
                 733
have: tlass nace, dtype: iota4
 quantidade_por_class = 500
flouers\_df[=pd,concat([flovers\_df[flovers\_df['class\_name']="i"]:=validade\_por\_cJ \ ass] \ for \ i \ in \ CLASS\_WE5])
f1ouers\_df.cIass\_naae.va1ue\_counts()
dBls}
dandelion
                586
```

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

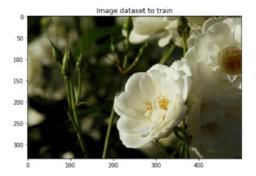
```
In [16]:
    X = flowers_df['filepath']
    y = flowers_df['label']

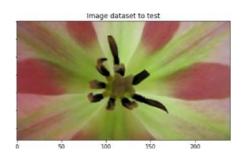
x_train, x_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=101)
```

Question-3:

Create Model

Solution:



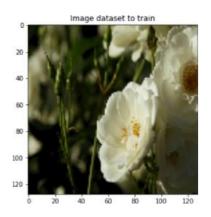


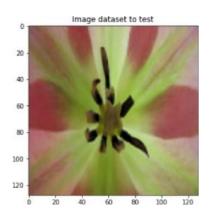
```
netunns a image that is reshaped and normalized
image = IE.c a st image, IF. Float32)
image = tf.image.resize(image, IMAGE_SIZE)
return image, Zabe1

train_data_norm = train_data_img.map(prepnocessing)

fig, ax = plt.subplots(1,2, figsize = (15,5))
```

```
fig, ax = plt.subplots(1,2, figsize = (15,5))
for i,1 iM train_data_norm.take(1):
    ax[0].set_title("Image dataset to train");
    ax[0].imshow(i);
for i,1 in test_data_norm.take(I):
    ax[1].set_title('Image dataset to test');
    ax[1].imshow(i);
```





```
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)
Te sl Data Shape (3 2, 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.Platten())
    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:

Compile The Model

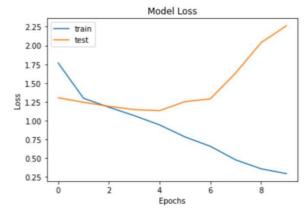
Question-6:

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 [=====
                  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
                      :========] - 40s 719ms/step - loss: 0.4778 - accuracy: 0.8257 - val_loss: 1.6341 - val_accuracy: 0.5080
      55/55 [====
      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
                  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)
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,
```

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()
```

