Assignment -3

Python Programming

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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]: !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()
          0
```

```
50 -

100 -

150 -

200 -

250 -

0 50 100 150 200 250 300
```

```
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]: 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...
                                                                      0
                                                             daisy
                                                                      0
           1 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                             daisy
           2 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                                      0
                                                             daisy
          3 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                             daisy
           4 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                             daisy
                                                                      0
In [12]: flowers_df.class_name.value_counts()
          dandelion 1052
Out[12]:
          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()

100 -150 -200 -250 -300 -

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)
```

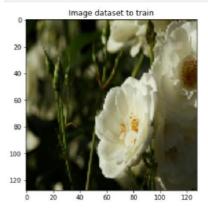
Question-3:

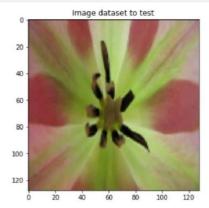
Create Model

Solution:



```
100 - 100 - 50 100 150 200
```





```
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.Platten())
    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='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
Fotal params: 3,480,306 Frainable params: 3,480,306 Non-trainable params: 0		

Question-5:

Compile The Model

Compile The Model

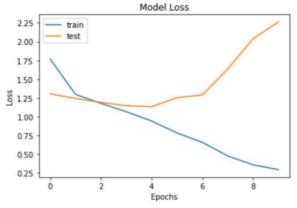
Solution:

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))
        55/55 [====
Epoch 2/10
                           =========] - 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
        55/55 [====
        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 [====
Epoch 5/10
                              :========] - 36s 650ms/step - loss: 1.0667 - accuracy: 0.5526 - val_loss: 1.1468 - val_accuracy: 0.5453
        55/55 [=====
Epoch 6/10
55/55 [=====
Epoch 7/10
55/55 [=====
                              =========] - 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
                              =========] - 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 [====:
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

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, 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()
```

