Assignment -1

Python Programming

Assignment Date	9 october 2022
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Student Roll Number	111519104044
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]:
            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 -

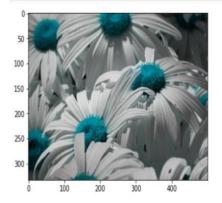
250 -

0 50 100 150 200 250 300
```

```
In [10]:
            GCS_PATH = "/content/drive/MyDrive/Flowers-Dataset/flowers"
            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...
                                                                  0
          1 /content/drive/MyDrive/Flowers-Dataset/flowers...
          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...
                                                                  0
                                                         daisy
In [12]: flowers_df.class_name.value_counts()
Out[12]: dandelion
                      1052
          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()
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)
```

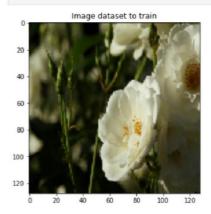
Question-3:

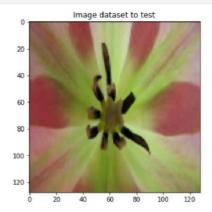
Create Model

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

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 124, 124, 6)	1000000
<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

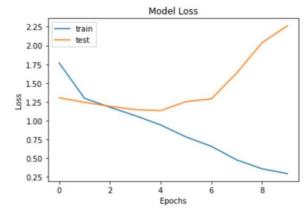
Question-6:

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
                     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
                  =========================== - 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, 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()
```

