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
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()
```

```
100 -

150 -

200 -

250 -

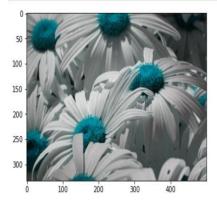
0 50 100 150 200 250 300
```

```
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
           array(['daisy', 'rose', 'dandelion', 'sunflower', 'tulip'], dtype='<U9')</pre>
 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...
                                                                   0
          2 /content/drive/MyDrive/Flowers-Dataset/flowers...
                                                                   0
                                                           daisy
          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()
          dandelion
                       1052
Out[12]:
          tulip
                        984
                        784
          rose
          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()
                        500
          daisy
Out[14]:
                        500
          rose
          dandelion
                       500
          sunflower
                       500
          tulip
```

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)
```

Tn [17].

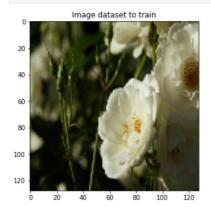
Question-3:

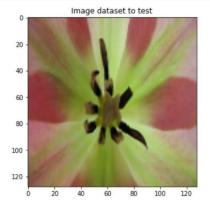
Create Model

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, 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)

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

Trainable params: 3,480,306 Non-trainable params: 0

Compile The Model

Question-5:

Compile The

Model

Solution:

Question-6:

Fit The Model

Solution:

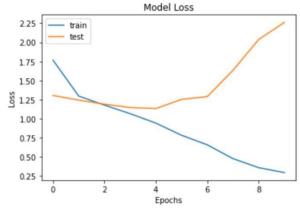
Fit The Model

```
In [26]: history = LeNet.fit(train_batches, epochs=10,batch_size = 16,validation_data=(test_batches))
       Epoch 1/10
       55/55 [====
Epoch 2/10
55/55 [====
Epoch 3/10
                       55/55 [====:
                     Epoch 4/10

55/55 [=========] - 36s 650ms/step - loss: 1.0667 - accuracy: 0.5526 - val_loss: 1.1468 - val_accuracy: 0.5528

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
55/55 [====
                         =========] - 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 10/10
       5/55 [===============] - 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)
          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()
```

