Assignment – 2

Flower Classification using CNN

Assignment Date	01 October 2022
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Maximum Marks	2 Marks

Task - 1: Import the necessary libraries

```
1. Import the necessary libraries

In [1]: import splitfolders import numpy as np import tensorflow as tf from tensorflow. keras.preprocessing.image import ImageDataGenerator from tensorflow.keras.preprocessing import image from tensorflow.keras import layers from tensorflow.keras.models import Sequential from tensorflow.keras.models import load_model from tensorflow.keras.layers import Dense,Convolution2D,MaxPooling2D,Flatten from tensorflow.keras.applications.resnet50 import preprocess_input, decode_predictions from tensorflow.keras.preprocessing import image import matplotlib.pyplot as plt
```

Task - 2: Download the dataset and perform image augmentation

2. Download the dataset and perform image augmentation In [2]: train_datagen = ImageDataGenerator(rescale=1./255,zoom_range=0.2,horizontal_flip=True,vertical_flip=False) In [3]: test_datagen = ImageDataGenerator(rescale=1./255) In [4]: input_folder = 'C:\\Users\\kumar\\OneDrive\\Documents\\IBM\\assignment_3\\flowers' In [5]: splitfolders.ratio(input_folder,output="C:\\Users\\kumar\\OneDrive\\Documents\\IBM\\assignment_3\\flowers\\flower

Task - 3: Create the model

Model should contain a Convolution Layer, MaxPooling Layer, Flatten Layer, Dense Layer and Output layer

3. Create the model

Model should contain a Convolution Layer, MaxPooling Layer, Flatten Layer, Dense Layer and Output layer

```
In [9]: model=Sequential()
model.add(Convolution2D(32,(3,3),input_shape=(64,64,3),activation='relu'))
model.add(MaxPooling2D(pool_size=(2,2)))
model.add(Flatten())
model.add(Dense(300,activation='relu'))
model.add(Dense(150,activation='relu'))
```

In [10]: model.summary()

Model: "sequential"

Layer (type)	Output	Shape	Param #
conv2d (Conv2D)	(None,	62, 62, 32)	896
max_pooling2d (MaxPooling2D)	(None,	31, 31, 32)	0
flatten (Flatten)	(None,	30752)	0
dense (Dense)	(None,	300)	9225900
dense_1 (Dense)	(None,	150)	45150

Total params: 9,271,946 Trainable params: 9,271,946 Non-trainable params: 0

In [11]: model.add(Dense(5,activation='softmax'))
 model.summary()

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 62, 62, 32)	896
max_pooling2d (MaxPooling2D)	(None, 31, 31, 32)	0
flatten (Flatten)	(None, 30752)	0
dense (Dense)	(None, 300)	9225900
dense_1 (Dense)	(None, 150)	45150
dense_2 (Dense)	(None, 5)	755

Total params: 9,272,701 Trainable params: 9,272,701 Non-trainable params: 0

Task – 4: Compile the model and train it for 20 epochs using model.fit() method

```
In [13]: epoch=20
       history = model.fit(x_train,steps_per_epoch=len(x_train),validation_data=x_test,validation_steps=len(x_test),epochs=epoch)
       Epoch 1/20
       144/144 [=
                            ========] - 44s 158ms/step - loss: 1.2757 - accuracy: 0.4676 - val_loss: 1.1261 - val_accuracy: 0.
       5549
       Epoch 2/20
                            =========] - 8s 57ms/step - loss: 1.0409 - accuracy: 0.5855 - val loss: 1.1483 - val accuracy: 0.57
       144/144 [=
       144/144 [==
                           ========] - 7s 52ms/step - loss: 0.9904 - accuracy: 0.6162 - val_loss: 1.1175 - val_accuracy: 0.57
       Epoch 4/20
       144/144 [==
                           =========] - 7s 49ms/step - loss: 0.8925 - accuracy: 0.6541 - val_loss: 1.0778 - val_accuracy: 0.58
                          ========] - 7s 51ms/step - loss: 0.8504 - accuracy: 0.6715 - val loss: 1.0019 - val accuracy: 0.63
       144/144 [=
       93
                       :==========] - 7s 51ms/step - loss: 0.8065 - accuracy: 0.6897 - val_loss: 1.0034 - val_accuracy: 0.63
       93
       Fnoch 7/20
                         ===========] - 7s 50ms/step - loss: 0.7544 - accuracy: 0.7123 - val_loss: 1.1677 - val_accuracy: 0.59
       144/144 [==
       Epoch 8/20
       144/144 [=======] - 7s 51ms/step - loss: 0.7005 - accuracy: 0.7303 - val_loss: 1.0580 - val_accuracy: 0.63
       12
       Epoch 9/20
       144/144 [==
                           =========] - 7s 51ms/step - loss: 0.6536 - accuracy: 0.7497 - val_loss: 1.0528 - val_accuracy: 0.63
       Epoch 10/20
       144/144 [===
                       :==========] - 8s 56ms/step - loss: 0.5679 - accuracy: 0.7810 - val_loss: 1.1217 - val_accuracy: 0.63
       Epoch 12/20
       144/144 [===:
                       Epoch 13/20
       144/144 [===
                       ========== ] - 7s 51ms/step - loss: 0.4846 - accuracy: 0.8166 - val_loss: 1.1872 - val_accuracy: 0.60
       81
       Epoch 14/20
                        :=========] - 7s 51ms/step - loss: 0.4804 - accuracy: 0.8236 - val_loss: 1.1117 - val_accuracy: 0.67
       51
     Epoch 15/20
     144/144 [===
                        :========] - 7s 51ms/step - loss: 0.4333 - accuracy: 0.8453 - val_loss: 1.1388 - val_accuracy: 0.67
     Epoch 16/20
     144/144 [===
                         ========] - 7s 51ms/step - loss: 0.4119 - accuracy: 0.8508 - val_loss: 1.1882 - val_accuracy: 0.66
     94
     Epoch 17/20
                      :========] - 7s 51ms/step - loss: 0.3981 - accuracy: 0.8552 - val_loss: 1.3760 - val_accuracy: 0.65
     144/144 [===
     144/144 [==:
                       43
     Fnoch 19/20
     144/144 [========= ] - 10s 66ms/step - loss: 0.3341 - accuracy: 0.8763 - val loss: 1.4009 - val accuracy: 0.6
     144/144 [===
                     :==========] - 8s 52ms/step - loss: 0.2888 - accuracy: 0.9053 - val_loss: 1.4764 - val_accuracy: 0.63
     91
```

Task – 5: Plot the training and validation accuracy along with training and validation loss

Task – 6: Save the model

6. Save the Model

```
In [17]: model.save('flowers.h5')
```

Task – 7: Test the model

7. Test the model

```
In [18]: img=image.load_img(r"C:\Users\kumar\OneDrive\Documents\IBM\assignment_3\flowers\flowersdataset\test\daisy\3706420943_66f3214862_n.
x=image.img_to_array(img)
x=np.expand_dims(x,axis=0)
y=np.argmax(model.predict(x),axis=1)
x_train.class_indices
index=['daisy', 'dandellion', 'rose', 'sunflower', 'tulip']
index[y[0]]

Out[18]: 'daisy'
```