Assignment -3

Assignment Date Student Name Student Roll Number Maximum Marks 01 October 2022 HEMALATHA J 111519104045 2 Marks

Question:1

Download the Dataset

Question-2:

Image Augmentation

Solution:

data = ImageDataGenerator(rescale = 1.0/225, zoom_range = 0.2, horizontal_flip = True, vertical_flip = False, validation_split=0.25)

train_data = data.flow_from_directory('/content/flowers/', target_size=(224,224), class_mode = 'categorical', subset= 'training')

test_data = data.flow_from_directory('/content/flowers',target_size=(224,224),class_mode = 'categorical', subset = 'validation')

train_data.class_indices

Image Augmentation

Question-3:

Create Model

Solution:

datamodel = Sequential()

Create Model

```
In [27]: datamodel = Sequential()
```

Question-4:

Add Layers

Solution:

```
datamodel.add(Convolution2D(32,(3,3),input_shape=(224,224,3),activation='relu')) datamodel.add(MaxPooling2D(pool_size=(2,2))) datamodel.add(Flatten()) datamodel.add(Dense(300,activation='relu')) datamodel.add(Dense(150,activation='relu')) datamodel.add(Dense(5,activation='softmax'))
```

Add Layers

```
In [28]:
    datamodel.add(Convolution2D(32,(3,3),input_shape=(224,224,3),activation='relu'))
    datamodel.add(MaxPooling2D|(pool_size=(2,2)))
    datamodel.add(Flatten())
    datamodel.add(Dense(300,activation='relu'))
    datamodel.add(Dense(150,activation='relu'))
    datamodel.add(Dense(5,activation='relu'))
```

Question-5:

Compile The Model

Soltion:

datamodel.compile(loss='categorical_crossentropy',optimizer='adam',metrics=['accuracy'])

Compile The Model

```
In [19]: datamodel.compile(loss='categorical_crossentropy',optimizer='adam',metrics=['accuracy'])
```

Question -6:

Fit The Model

Solution:

datamodel.fit(train_data,steps_per_epoch=len(train_data),validation_data=test_data,validation_steps=len(test_data),epochs=15)

```
In [20]: datamodel.fit(train_data,steps_per_epoch=len(train_data),validation_data=test_data,validation_steps=len(test_data),e
         Epoch 1/15
          102/102 [=
                                                ===] - 275s 3s/step - loss: 7.5656 - accuracy: 0.4166 - val loss: 1.2590 - val
          accuracy: 0.5199
          Epoch 2/15
          102/102 [=
                                                    - 267s 3s/step - loss: 1.1255 - accuracy: 0.5469 - val_loss: 1.1373 - val_
          accuracy: 0.5653
          Epoch 3/15
          102/102 [=
                                                      269s 3s/step - loss: 1.0337 - accuracy: 0.5985 - val_loss: 1.0682 - val_
         accuracy: 0.6033
Epoch 4/15
                                                      262s 3s/step - loss: 0.9246 - accuracy: 0.6445 - val_loss: 1.0091 - val_
          102/102 [=
          accuracy: 0.6089
          Epoch 5/15
          102/102 [==
                                                      282s 3s/step - loss: 0.8749 - accuracy: 0.6683 - val_loss: 1.0787 - val_
          accuracy: 0.5987
          Epoch 6/15
          102/102 [=
                                                      265s 3s/step - loss: 0.8034 - accuracy: 0.7082 - val_loss: 1.0256 - val_
          accuracy: 0.6163
          102/102 [==
                                                    - 271s 3s/step - loss: 0.7703 - accuracy: 0.7106 - val_loss: 1.0172 - val_
          accuracy: 0.6293
          Epoch 8/15
          102/102 [==
                                              ====] - 278s 3s/step - loss: 0.7541 - accuracy: 0.7103 - val_loss: 1.0860 - val_
          accuracy: 0.6033
          Epoch 9/15
         102/102 [======
accuracy: 0.6330
                                                    - 276s 3s/step - loss: 0.7013 - accuracy: 0.7415 - val loss: 1.0853 - val
          Epoch 10/15
          102/102 [======
accuracy: 0.6367
                                           ======] - 267s 3s/step - loss: 0.6857 - accuracy: 0.7409 - val_loss: 1.0300 - val_
          Epoch 11/15
          102/102 [======
                                     :=======] - 273s 3s/step - loss: 0.6601 - accuracy: 0.7477 - val loss: 0.9765 - val
          Epoch 10/15
          102/102 [=
                                                     - 267s 3s/step - loss: 0.6857 - accuracy: 0.7409 - val_loss: 1.0300 - val_
          accuracy: 0.6367
Epoch 11/15
          102/102 [=
                                                     - 273s 3s/step - loss: 0.6601 - accuracy: 0.7477 - val_loss: 0.9765 - val_
          accuracy: 0.6589
Epoch 12/15
          102/102 [==:
                                                     - 267s 3s/step - loss: 0.5935 - accuracy: 0.7795 - val_loss: 1.0453 - val_
          accuracy: 0.6497
          102/102 [==:
                                                     - 271s 3s/step - loss: 0.5759 - accuracy: 0.7928 - val_loss: 1.0114 - val_
          accuracy: 0.6701
          Epoch 14/15
          102/102 [======
accuracy: 0.6432
                                          =======] - 269s 3s/step - loss: 0.5412 - accuracy: 0.8027 - val_loss: 1.0206 - val_
          Epoch 15/15
          102/102 [==:
                                      ========] - 267s 3s/step - loss: 0.5327 - accuracy: 0.8039 - val_loss: 1.1359 - val_
          accuracy: 0.6395
 Out[20]: <keras.callbacks.History at 0x7f73674bcf90>
```

Save The Model

Save The Model

```
In [21]: datamodel.save('flowers.h5')
```

Question-7:Save the model

solution:

datamodel.save('flowers.h5')

Question-8:

Test The Model **Solution**:

```
img =
```

image.load_img('/content/flowers/dandelion/10437652486_aa86c14985.jpg',target_size=(224,224))
temp_arr = image.img_to_array(img)
dim = np.expand_dims(temp_arr,axis=0)
temp=np.argmax(datamodel.predict(dim),axis=1)

$index = ['Daisy', 'Dandelion', 'Rose', 'Sunflower', 'Tulip'] \\ index[temp[0]]$

Test The Model

```
In [23]: img = image.load_img('/content/flowers/dandelion/10437652486_aa86c14985.jpg', target_size=(224,224))

Out[23]:

In [24]: temp_arr = image.img_to_array(img)

In [25]: dim = np.expand_dims(temp_arr,axis=0)
    temp=np.argmax(datamodel.predict(dim),axis=1)
    index = ['Daisy', 'Dandelion', 'Rose', 'Sunflower', 'Tulip']
    index[temp[0]]
Out[25]: 'Daisy'
```