## **Assignment-3**

Assignment Date	8 October 2022
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Students Register Number	820319106001
Maximum marks	2 marks

```
import tensorflow as tf
In [7]:
from keras.preprocessing.image import ImageDataGenerator
In []:
#Augmenting the input training images
In [11]:
train_datagen = ImageDataGenerator(
        rescale=1./255,
        shear_range=0.2,
        zoom_range=0.2,
        horizontal_flip=True)
training_set = train_datagen.flow_from_directory(
        'training',
        target size=(64, 64),
        batch size=32,
        class mode='categorical')
Found 4103 images belonging to 5 classes.
```

```
In [12]:
test datagen = ImageDataGenerator(
        rescale=1./255)
test data = test datagen.flow from directory(
        'Testing',
        target_size=(64, 64),
        batch size=32,
        class mode='categorical')
Found 214 images belonging to 5 classes.
In []:
#Building the model
In [13]:
cnn = tf.keras.models.Sequential()
In []:
#Adding convolution layer
In [14]:
cnn.add(tf.keras.layers.Conv2D(filters=64,kernel size=3,activation
="relu", input shape =[64,64,3]))
cnn.add(tf.keras.layers.MaxPool2D(pool size = 2,strides=2))
In [15]:
cnn.add(tf.keras.layers.Conv2D(filters=64,kernel size=3,activation ="relu"))
cnn.add(tf.keras.layers.MaxPool2D(pool size = 2,strides=2))
In [16]:
```

```
cnn.add(tf.keras.layers.Dropout(0.5))
In []:
# Flattening the layers
In [17]:
cnn.add(tf.keras.layers.Flatten())
In []:
# Adding dense layers(Hidden Layers)
In [18]:
cnn.add(tf.keras.layers.Dense(units=128 ,activation ="relu"))
In [19]:
cnn.add(tf.keras.layers.Dense(units=5,activation="softmax"))
In []:
#compilation of the neural network model
In [20]:
cnn.compile(optimizer="rmsprop",loss="categorical_crossentropy" ,metrics
=["accuracy"])
In []:
#Fitting the neural network model and training it
In [41]:
cnn.fit(x = training set , validation data = test data , epochs = 30 )
Epoch 1/30
accuracy: 0.4350 - val loss: 1.0596 - val accuracy: 0.6168
Epoch 2/30
```

```
129/129 [============= ] - 33s 253ms/step - loss: 1.0957 -
accuracy: 0.5659 - val loss: 1.1546 - val accuracy: 0.6168
Epoch 3/30
129/129 [============= - 36s 279ms/step - loss: 0.9823 -
accuracy: 0.6176 - val loss: 1.0383 - val accuracy: 0.5841
Epoch 4/30
129/129 [============= ] - 37s 285ms/step - loss: 0.9194 -
accuracy: 0.6432 - val loss: 0.8612 - val accuracy: 0.6776
Epoch 5/30
accuracy: 0.6727 - val loss: 1.1994 - val accuracy: 0.5514
Epoch 6/30
accuracy: 0.6856 - val loss: 0.9825 - val accuracy: 0.6916
Epoch 7/30
accuracy: 0.7002 - val loss: 0.9143 - val accuracy: 0.6636
Epoch 8/30
129/129 [============ ] - 36s 280ms/step - loss: 0.7603 -
accuracy: 0.7090 - val loss: 0.8084 - val accuracy: 0.7243
Epoch 9/30
accuracy: 0.7187 - val loss: 0.8042 - val accuracy: 0.7150
Epoch 10/30
accuracy: 0.7387 - val loss: 0.9286 - val accuracy: 0.6589
Epoch 11/30
129/129 [============ ] - 35s 273ms/step - loss: 0.6722 -
accuracy: 0.7453 - val loss: 1.0362 - val accuracy: 0.6822
```

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Epoch 12/30
129/129 [============= ] - 35s 270ms/step - loss: 0.6659 -
accuracy: 0.7534 - val loss: 0.7733 - val accuracy: 0.7056
Epoch 13/30
accuracy: 0.7655 - val loss: 0.8955 - val accuracy: 0.6916
Epoch 14/30
accuracy: 0.7702 - val loss: 0.9361 - val accuracy: 0.6542
Epoch 15/30
129/129 [============ - 36s 279ms/step - loss: 0.5988 -
accuracy: 0.7780 - val loss: 0.8789 - val accuracy: 0.6916
Epoch 16/30
129/129 [============ ] - 36s 281ms/step - loss: 0.5822 -
accuracy: 0.7775 - val loss: 0.9812 - val accuracy: 0.6729
Epoch 17/30
accuracy: 0.7870 - val loss: 0.8973 - val accuracy: 0.7056
Epoch 18/30
accuracy: 0.7875 - val loss: 0.8542 - val accuracy: 0.7056
Epoch 19/30
129/129 [============ ] - 39s 305ms/step - loss: 0.5624 -
accuracy: 0.7955 - val loss: 0.7468 - val accuracy: 0.7430
Epoch 20/30
129/129 [=========== ] - 39s 303ms/step - loss: 0.5542 -
accuracy: 0.7919 - val loss: 0.8988 - val accuracy: 0.7150
Epoch 21/30
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129/129 [============= ] - 43s 329ms/step - loss: 0.5241 -
accuracy: 0.8040 - val loss: 1.0677 - val accuracy: 0.6963
Epoch 22/30
accuracy: 0.8172 - val loss: 0.8774 - val accuracy: 0.7243
Epoch 23/30
accuracy: 0.8172 - val loss: 0.8348 - val accuracy: 0.6963
Epoch 24/30
accuracy: 0.8153 - val loss: 0.9380 - val accuracy: 0.6916
Epoch 25/30
accuracy: 0.8284 - val loss: 0.9572 - val accuracy: 0.7056
Epoch 26/30
accuracy: 0.8360 - val loss: 0.8506 - val accuracy: 0.7056
Epoch 27/30
129/129 [=========== ] - 39s 302ms/step - loss: 0.4734 -
accuracy: 0.8216 - val loss: 1.2935 - val accuracy: 0.6168
Epoch 28/30
accuracy: 0.8272 - val loss: 0.8751 - val accuracy: 0.6869
Epoch 29/30
129/129 [============ ] - 37s 290ms/step - loss: 0.4375 -
accuracy: 0.8372 - val loss: 0.9651 - val accuracy: 0.6729
Epoch 30/30
129/129 [============ ] - 39s 299ms/step - loss: 0.4292 -
accuracy: 0.8501 - val loss: 1.0778 - val accuracy: 0.6963
```

```
Out[41]:
<keras.callbacks.History at 0x2bf28ab59b0>
In [42]:
cnn.fit(x = training set , validation data = test data , epochs = 30)
Epoch 1/30
accuracy: 0.8496 - val loss: 0.9867 - val accuracy: 0.6729
Epoch 2/30
accuracy: 0.8469 - val_loss: 1.0115 - val_accuracy: 0.7056
Epoch 3/30
accuracy: 0.8550 - val loss: 0.8851 - val accuracy: 0.7150
Epoch 4/30
129/129 [============ ] - 44s 341ms/step - loss: 0.4077 -
accuracy: 0.8513 - val_loss: 1.1110 - val_accuracy: 0.6916
Epoch 5/30
accuracy: 0.8603 - val loss: 1.2546 - val accuracy: 0.7103
Epoch 6/30
129/129 [============ - 42s 327ms/step - loss: 0.4018 -
accuracy: 0.8630 - val loss: 0.9946 - val accuracy: 0.6916
Epoch 7/30
accuracy: 0.8640 - val loss: 1.0004 - val accuracy: 0.7243
```

accuracy: 0.8655 - val\_loss: 1.0725 - val\_accuracy: 0.6916

Epoch 8/30

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Epoch 9/30
accuracy: 0.8582 - val loss: 1.0544 - val accuracy: 0.6916
Epoch 10/30
accuracy: 0.8652 - val loss: 0.9719 - val accuracy: 0.6963
Epoch 11/30
accuracy: 0.8686 - val loss: 0.9270 - val accuracy: 0.7336
Epoch 12/30
accuracy: 0.8647 - val loss: 0.9987 - val accuracy: 0.7196
Epoch 13/30
accuracy: 0.8718 - val loss: 0.8642 - val accuracy: 0.7196
Epoch 14/30
accuracy: 0.8786 - val loss: 1.1820 - val accuracy: 0.6822
Epoch 15/30
accuracy: 0.8762 - val loss: 1.0773 - val accuracy: 0.7150
Epoch 16/30
accuracy: 0.8852 - val loss: 1.3577 - val accuracy: 0.7009
Epoch 17/30
accuracy: 0.8796 - val loss: 1.0770 - val accuracy: 0.7150
Epoch 18/30
```

```
129/129 [============= ] - 63s 477ms/step - loss: 0.3444 -
accuracy: 0.8755 - val loss: 0.9273 - val accuracy: 0.7243
Epoch 19/30
accuracy: 0.8835 - val loss: 1.1471 - val accuracy: 0.6776
Epoch 20/30
accuracy: 0.8869 - val loss: 1.1275 - val accuracy: 0.7103
Epoch 21/30
accuracy: 0.8864 - val loss: 1.2780 - val accuracy: 0.6963
Epoch 22/30
accuracy: 0.8867 - val loss: 1.0580 - val accuracy: 0.7056
Epoch 23/30
accuracy: 0.8903 - val loss: 1.2799 - val accuracy: 0.7383
Epoch 24/30
129/129 [============ ] - 101s 785ms/step - loss: 0.3164 -
accuracy: 0.8884 - val loss: 1.3724 - val accuracy: 0.7056
Epoch 25/30
accuracy: 0.8945 - val loss: 1.2431 - val accuracy: 0.7009
Epoch 26/30
accuracy: 0.8945 - val loss: 0.9750 - val accuracy: 0.7056
Epoch 27/30
129/129 [============ ] - 111s 851ms/step - loss: 0.3087 -
accuracy: 0.9020 - val loss: 1.4106 - val accuracy: 0.7056
```

```
Epoch 28/30
accuracy: 0.8935 - val loss: 0.9878 - val accuracy: 0.7243
Epoch 29/30
accuracy: 0.8976 - val loss: 1.1608 - val accuracy: 0.6963
Epoch 30/30
accuracy: 0.8913 - val_loss: 1.4083 - val_accuracy: 0.7336
Out[42]:
<keras.callbacks.History at 0x2bf223fcfd0>
In []:
#preprocess the test image
In [43]:
import numpy as np
In [55]:
image
tf.keras.preprocessing.image.load_img("prediction/tu.jpg",target_size=(64,64)
input arr = tf.keras.preprocessing.image.img to array(image)
input arr = np.expand dims(input arr,axis=0)
result = cnn.predict(input arr)
1/1 [======] - 0s 79ms/step
In [52]:
training set.class indices
Out[52]:
```

```
{'Daisy': 0, 'Dandelion': 1, 'rose': 2, 'sunflower': 3, 'tulip': 4}
In [56]:
print(result)
[[0. 0. 0. 0. 1.]]
In [ ]:
#Mapping the result to the values
In [57]:
if result[0][0] == 1:
   print("daisy")
elif result[0][1] == 1:
    print("dandelion")
elif result[0][2] == 1:
    print("rose")
elif result[0][3] ==1:
   print("suflower")
elif result[0][4] == 1:
   print("tulip")
tulip
```

In []: