

### Assignment -3

Problem Statement:- Build CNN Model for Classification Of Flowers

Assignment Date	06 October 2022
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Maximum Marks	2 Marks

#### 1. Download the Dataset

Importing all necessary libraries

Solution:

```
!pip install split-folders
```

```
import splitfolders
```

```
import numpy as np
```

```
import tensorflow as tf
```

Load dataset

Solution:

```
import zipfile
```

```
from google.colab import drive
drive.mount('/content/drive')
!unzip /content/drive/My Drive/Flowers-Dataset.zip
```

Output:

```
Drive already mounted at /content/drive; to attempt to forcibly remount, call
drive.mount("/content/drive", force_remount=True).
Archive: /content/drive/My Drive/Flowers-Dataset.zip
replace flowers/daisy/100080576_f52e8ee070_n.jpg? [y]es, [n]o, [A]ll, [N]one,
[r]ename:
```

Split dataset into training data, validation data, testing data

Solution:

```
splitfolders.ratio("/content/flowers", output="/content/flowers", seed=1337,
ratio=(.8, .1, .1), group_prefix=None)
```

Output:

```
Copying files: 4317 files [00:01, 3694.23 files/s]
```

#### 2. Image Augmentation

Solution:

```
import keras
```

```
from tensorflow.keras.preprocessing.image import ImageDataGenerator
```

```
train_datagen=ImageDataGenerator(rescale=1./255, zoom_range=0.2, horizontal_flip=
True, vertical_flip=False)
```

```
test_datagen=ImageDataGenerator(rescale=1./255)

gentrain=train_datagen.flow_from_directory("/content/flowers/train",target_size
=(64,64),class_mode="categorical",batch_size=100)

gentest=test_datagen.flow_from_directory("/content/flowers/val",target_size=(64
,64),class_mode="categorical",batch_size=100)

genval=test_datagen.flow_from_directory("/content/flowers/test",target_size=(64
,64),class_mode="categorical",batch_size=100)
```

### 3. Create Model

Solution:

```
from tensorflow.keras.models import Sequential

from tensorflow.keras.layers import Convolution2D,MaxPooling2D,Flatten,Dense
model=Sequential()
```

### 4. Add Layers (Convolution,MaxPooling,Flatten,Dense-(Hidden Layers),Output)

Solution:

```
model.add(Convolution2D(32,(3,3),activation="relu",input_shape=(64,64,3)))

model.add(MaxPooling2D(pool_size=(2,2)))

model.add(Flatten())

model.add(Dense(300,kernel_initializer="random_uniform",activation="relu"))

model.add(Dense(5,kernel_initializer="random_uniform",activation="softmax"))
```

### 5. Compile The Model

Solution:

```
model.compile(loss="categorical_crossentropy",optimizer="adam",metrics=["accuracy"])
```

### 6. Fit The Model

Solution:

```
model.fit_generator(gentrain,steps_per_epoch=15,epochs=30,validation_data=gentest,validation_steps=10)
```

Output:

```
Epoch 1/30
15/15 [=====] - ETA: 0s - loss: 1.0158 - accuracy: 0.6027
15/15 [=====] - 13s 867ms/step - loss: 1.0158 - accuracy: 0.6027 - val_loss: 1.0514 - val_accuracy: 0.6093
Epoch 2/30
15/15 [=====] - 10s 665ms/step - loss: 0.9545 - accuracy: 0.6220
Epoch 3/30
15/15 [=====] - 10s 656ms/step - loss: 0.9348 - accuracy: 0.6347
Epoch 4/30
```

15/15 [=====] - 10s 675ms/step - loss: 0.9333 -  
accuracy: 0.6253  
Epoch 5/30  
15/15 [=====] - 11s 721ms/step - loss: 0.9238 -  
accuracy: 0.6460  
Epoch 6/30  
15/15 [=====] - 10s 644ms/step - loss: 0.8809 -  
accuracy: 0.6563  
Epoch 7/30  
15/15 [=====] - 10s 666ms/step - loss: 0.8613 -  
accuracy: 0.6715  
Epoch 8/30  
15/15 [=====] - 12s 739ms/step - loss: 0.8220 -  
accuracy: 0.6707  
Epoch 9/30  
15/15 [=====] - 10s 657ms/step - loss: 0.8181 -  
accuracy: 0.6933  
Epoch 10/30  
15/15 [=====] - 10s 664ms/step - loss: 0.8075 -  
accuracy: 0.6860  
Epoch 11/30  
15/15 [=====] - 10s 628ms/step - loss: 0.7976 -  
accuracy: 0.6935  
Epoch 12/30  
15/15 [=====] - 10s 640ms/step - loss: 0.7626 -  
accuracy: 0.7073  
Epoch 13/30  
15/15 [=====] - 10s 641ms/step - loss: 0.7525 -  
accuracy: 0.7039  
Epoch 14/30  
15/15 [=====] - 10s 650ms/step - loss: 0.7359 -  
accuracy: 0.7153  
Epoch 15/30  
15/15 [=====] - 10s 659ms/step - loss: 0.7489 -  
accuracy: 0.7127  
Epoch 16/30  
15/15 [=====] - 10s 633ms/step - loss: 0.7419 -  
accuracy: 0.7190  
Epoch 17/30  
15/15 [=====] - 10s 648ms/step - loss: 0.7459 -  
accuracy: 0.7140  
Epoch 18/30  
15/15 [=====] - 10s 631ms/step - loss: 0.7001 -  
accuracy: 0.7410  
Epoch 19/30  
15/15 [=====] - 10s 632ms/step - loss: 0.6900 -  
accuracy: 0.7376  
Epoch 20/30  
15/15 [=====] - 10s 659ms/step - loss: 0.6840 -  
accuracy: 0.7487  
Epoch 21/30  
15/15 [=====] - 10s 668ms/step - loss: 0.6905 -  
accuracy: 0.7367  
Epoch 22/30  
15/15 [=====] - 10s 635ms/step - loss: 0.6613 -  
accuracy: 0.7493  
Epoch 23/30  
15/15 [=====] - 10s 659ms/step - loss: 0.6479 -  
accuracy: 0.7453  
Epoch 24/30  
15/15 [=====] - 10s 621ms/step - loss: 0.6366 -  
accuracy: 0.7672  
Epoch 25/30  
15/15 [=====] - 10s 644ms/step - loss: 0.6610 -  
accuracy: 0.7541  
Epoch 26/30

```
15/15 [=====] - 10s 656ms/step - loss: 0.6416 -  
accuracy: 0.7627  
Epoch 27/30  
15/15 [=====] - 12s 772ms/step - loss: 0.5998 -  
accuracy: 0.7803  
Epoch 28/30  
15/15 [=====] - 10s 658ms/step - loss: 0.5547 -  
accuracy: 0.7887  
Epoch 29/30  
15/15 [=====] - 10s 679ms/step - loss: 0.5501 -  
accuracy: 0.7840  
Epoch 30/30  
15/15 [=====] - 10s 672ms/step - loss: 0.5859 -  
accuracy: 0.7833
```

## 7. Save The Model

Solution:

```
model.save("./flower.h5")
```

## 8. Test The Model

Solution:

```
model.evaluate(genval)
```

Output:

```
5/5 [=====] - 2s 279ms/step - loss: 0.9311 - accuracy:  
0.6759  
[0.9310972094535828, 0.6758620738983154]
```