

Assignment -3

Problem Statement:- Build CNN Model for Classification Of Flowers

Assignment Date	19 September 2022
Student Name	Ms. Nandhini.R
Student Roll Number	2019504552
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=genval, 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 - accu
racy: 0.7367
Epoch 22/30
15/15 [=====] - 10s 635ms/step - loss: 0.6613 - accu
racy: 0.7493
Epoch 23/30
15/15 [=====] - 10s 659ms/step - loss: 0.6479 - accu
racy: 0.7453
Epoch 24/30
15/15 [=====] - 10s 621ms/step - loss: 0.6366 - accu
racy: 0.7672
Epoch 25/30
15/15 [=====] - 10s 644ms/step - loss: 0.6610 - accu
racy: 0.7541
Epoch 26/30
15/15 [=====] - 10s 656ms/step - loss: 0.6416 - accu
racy: 0.7627
Epoch 27/30
15/15 [=====] - 12s 772ms/step - loss: 0.5998 - accu
racy: 0.7803
Epoch 28/30
15/15 [=====] - 10s 658ms/step - loss: 0.5547 - accu
racy: 0.7887
Epoch 29/30
15/15 [=====] - 10s 679ms/step - loss: 0.5501 - accu
racy: 0.7840
Epoch 30/30
15/15 [=====] - 10s 672ms/step - loss: 0.5859 - accu
racy: 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 - accurac
y: 0.6759
[0.9310972094535828, 0.6758620738983154]
```