

# Assignment – 3

## Python Programming

Assignment Date	03/10/2022
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Maximum Mark	2 Mark

Problem Statement : Build CNN Model for Classification Of Flowers

### 1.Download the dataset

```
In [ ]: #unzip the dataset
!unzip Flowers-Dataset.zip

Archive:  Flowers-Dataset.zip
  inflating: flowers/daisy/100080576_f52e8ee070_n.jpg
  inflating: flowers/daisy/10140303196_b88d3d6cec.jpg
  inflating: flowers/daisy/10172379554_b296050f82_n.jpg
  inflating: flowers/daisy/10172567486_2748826a8b.jpg
  inflating: flowers/daisy/10172636503_21bededa75_n.jpg
  inflating: flowers/daisy/102841525_bd6628ae3c.jpg
  inflating: flowers/daisy/10300722094_28fa978807_n.jpg
  inflating: flowers/daisy/1031799732_e7f4008c03.jpg
  inflating: flowers/daisy/10391248763_1d16681106_n.jpg
  inflating: flowers/daisy/10437754174_22ec990b77_m.jpg
  inflating: flowers/daisy/10437770546_80b6f7bdd3_m.jpg
  inflating: flowers/daisy/10437929963_bc13eebe0c.jpg
  inflating: flowers/daisy/10466290366_cc72e33532.jpg
  inflating: flowers/daisy/10466558316_a7198b87e2.jpg
  inflating: flowers/daisy/10555749515_13a12a026e.jpg
  inflating: flowers/daisy/10555815624_dc211569b0.jpg
  inflating: flowers/daisy/10555826524_423eb8bf71_n.jpg
  inflating: flowers/daisy/10559679065_50d2b16f6d.jpg
  inflating: flowers/daisy/105806915_a9c13e2106_n.jpg
  inflating: flowers/daisy/10712722853_5632165b04.jpg
  inflating: flowers/daisy/107592979_aaa9cdfef78_m.jpg
  inflating: flowers/daisy/10770585085_4742b9dac3_n.jpg
  inflating: flowers/daisy/10841136265_af473efc60.jpg
  inflating: flowers/daisy/10993710036_2033222c91.jpg
  inflating: flowers/daisy/10993818044_4c19b86c82.jpg
  inflating: flowers/daisy/10994032453_ac7f8d9e2e.jpg
  inflating: flowers/daisy/11023214096_b5b39fab08.jpg
  inflating: flowers/daisy/11023272144_fce94401f2_m.jpg
```

### 2. Image Augmentation

```
In [ ]: import numpy as np
import tensorflow as tf
from tensorflow.keras import layers
from tensorflow.keras.models import Sequential
import matplotlib.pyplot as plt

In [ ]: from tensorflow.keras.preprocessing.image import ImageDataGenerator

In [ ]: train_datagen = ImageDataGenerator(rescale = 1./255, horizontal_flip = True, vertical_flip = True, zoom_range = 0.2)

In [ ]: x_train = train_datagen.flow_from_directory(r"/content/flowers", target_size = (64,64) , class_mode = "categorical", batch_size = 100)

Found 4317 images belonging to 5 classes.

In [ ]: #Image Augmentation accuracy
data_augmentation = Sequential(
    [
        layers.RandomFlip("horizontal",input_shape=(img_height, img_width, 3)),
        layers.RandomRotation(0.1),
        layers.RandomZoom(0.1),
    ]
)
```

### 3.Create a model

```
In [ ]: from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Convolution2D,MaxPooling2D,Flatten,Dense
model = Sequential()
```

```
In [ ]: train_ds = tf.keras.utils.image_dataset_from_directory(
    data_dir,
    validation_split=0.2,
    subset="training",
    seed=123,
    image_size=(img_height, img_width),
    batch_size=batch_size)
```

Found 4317 files belonging to 5 classes.  
Using 3454 files for training.

```
In [ ]: val_ds = tf.keras.utils.image_dataset_from_directory(
    data_dir,
    validation_split=0.2,
    subset="validation",
    seed=123,
    image_size=(img_height, img_width),
    batch_size=batch_size)
```

Found 4317 files belonging to 5 classes.  
Using 863 files for validation.

```
In [ ]: class_names = train_ds.class_names
print(class_names)

['daisy', 'dandelion', 'rose', 'sunflower', 'tulip']
```

```
In [ ]: plt.figure(figsize=(10, 10))
for images, labels in train_ds.take(1):
    for i in range(9):
        ax = plt.subplot(3, 3, i + 1)
        plt.imshow(images[i].numpy().astype("uint8"))
        plt.title(class_names[labels[i]])
        plt.axis("off")
```



## Add layers (Convolution,MaxPooling,Flatten,Dense-(HiddenLayers),Output)

```
In [ ]: 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, activation = "relu"))
model.add(Dense(150, activation = "relu")) #multiple dense layers
model.add(Dense(5, activation = "softmax")) #output layer
```

```
In [ ]: num_classes = len(class_names)

model = Sequential([
    data_augmentation,
    layers.Rescaling(1./255, input_shape=(img_height, img_width, 3)),
    layers.Conv2D(16, 3, padding='same', activation='relu'),
    layers.MaxPooling2D(),
    layers.Conv2D(32, 3, padding='same', activation='relu'),
    layers.MaxPooling2D(),
    layers.Conv2D(64, 3, padding='same', activation='relu'),
    layers.MaxPooling2D(),
    layers.Flatten(),
    layers.Dense(128, activation='relu'),
    layers.Dense(num_classes)
])
```

## Compile The Model

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

Out[ ]: 44

```
In [ ]: model.compile(optimizer='adam',
                    loss=tf.keras.losses.SparseCategoricalCrossentropy(from_logits=True),
                    metrics=['accuracy'])

epochs=10
history = model.fit(
    train_ds,
    validation_data=val_ds,
    epochs=epochs
)
```

```
Epoch 1/10
108/108 [=====] - 132s 1s/step - loss: 1.2821 - accuracy: 0.4537 - val_loss: 1.0988 - val_accuracy: 0.5458
Epoch 2/10
108/108 [=====] - 130s 1s/step - loss: 1.0298 - accuracy: 0.5921 - val_loss: 0.9494 - val_accuracy: 0.6304
Epoch 3/10
108/108 [=====] - 129s 1s/step - loss: 0.9274 - accuracy: 0.6448 - val_loss: 0.9927 - val_accuracy: 0.6130
Epoch 4/10
108/108 [=====] - 129s 1s/step - loss: 0.9000 - accuracy: 0.6642 - val_loss: 0.9264 - val_accuracy: 0.6419
Epoch 5/10
108/108 [=====] - 136s 1s/step - loss: 0.8432 - accuracy: 0.6778 - val_loss: 0.8499 - val_accuracy: 0.6674
Epoch 6/10
108/108 [=====] - 130s 1s/step - loss: 0.8166 - accuracy: 0.6888 - val_loss: 0.8714 - val_accuracy: 0.6732
Epoch 7/10
108/108 [=====] - 130s 1s/step - loss: 0.7726 - accuracy: 0.7064 - val_loss: 0.7873 - val_accuracy: 0.6895
Epoch 8/10
108/108 [=====] - 130s 1s/step - loss: 0.7262 - accuracy: 0.7250 - val_loss: 0.7957 - val_accuracy: 0.6860
Epoch 9/10
108/108 [=====] - 128s 1s/step - loss: 0.7094 - accuracy: 0.7284 - val_loss: 0.7960 - val_accuracy: 0.7068
Epoch 10/10
108/108 [=====] - 130s 1s/step - loss: 0.6820 - accuracy: 0.7383 - val_loss: 0.7914 - val_accuracy: 0.6941
```

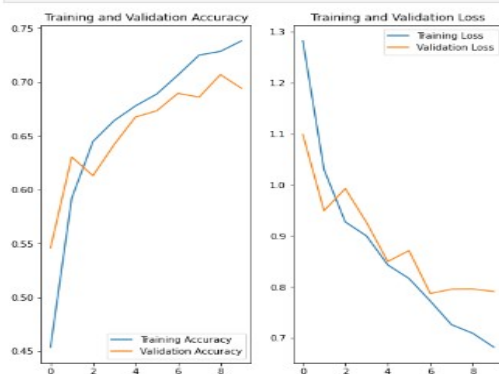
```
In [ ]: acc = history.history['accuracy']
val_acc = history.history['val_accuracy']

loss = history.history['loss']
val_loss = history.history['val_loss']

epochs_range = range(epochs)

plt.figure(figsize=(8, 8))
plt.subplot(1, 2, 1)
plt.plot(epochs_range, acc, label='Training Accuracy')
plt.plot(epochs_range, val_acc, label='Validation Accuracy')
plt.legend(loc='lower right')
plt.title('Training and Validation Accuracy')

plt.subplot(1, 2, 2)
plt.plot(epochs_range, loss, label='Training Loss')
plt.plot(epochs_range, val_loss, label='Validation Loss')
plt.legend(loc='upper right')
plt.title('Training and Validation Loss')
plt.show()
```



## Fit The Model

```
In [ ]: model.fit(x_train, epochs = 15, steps_per_epoch = len(x_train))

Epoch 1/15
44/44 [=====] - 31s 684ms/step - loss: 1.7914 - accuracy: 0.3588
Epoch 2/15
44/44 [=====] - 29s 648ms/step - loss: 1.1730 - accuracy: 0.5045
Epoch 3/15
44/44 [=====] - 29s 650ms/step - loss: 1.0967 - accuracy: 0.5529
Epoch 4/15
44/44 [=====] - 29s 648ms/step - loss: 1.0351 - accuracy: 0.5939
Epoch 5/15
44/44 [=====] - 29s 645ms/step - loss: 0.9920 - accuracy: 0.6127
Epoch 6/15
44/44 [=====] - 30s 677ms/step - loss: 0.9659 - accuracy: 0.6259
Epoch 7/15
44/44 [=====] - 29s 648ms/step - loss: 0.9129 - accuracy: 0.6426
Epoch 8/15
44/44 [=====] - 29s 647ms/step - loss: 0.9085 - accuracy: 0.6433
Epoch 9/15
44/44 [=====] - 32s 717ms/step - loss: 0.8597 - accuracy: 0.6620
Epoch 10/15
44/44 [=====] - 30s 674ms/step - loss: 0.8350 - accuracy: 0.6824
Epoch 11/15
44/44 [=====] - 29s 648ms/step - loss: 0.8420 - accuracy: 0.6718
Epoch 12/15
44/44 [=====] - 29s 650ms/step - loss: 0.7857 - accuracy: 0.7030
Epoch 13/15
44/44 [=====] - 29s 649ms/step - loss: 0.7868 - accuracy: 0.7000
Epoch 14/15
44/44 [=====] - 29s 650ms/step - loss: 0.7542 - accuracy: 0.7132
Epoch 15/15
44/44 [=====] - 30s 676ms/step - loss: 0.7467 - accuracy: 0.7107
Out[ ]: <keras.callbacks.History at 0x7f602ce90090>
```

## Save The Model

```
In [ ]: model.save("flowers.h1")

In [ ]: model.save("flowers.m5")#another model to show the accuracy
```

## Test The Model

```
In [ ]: from tensorflow.keras.models import load_model
        from tensorflow.keras.preprocessing import image
        import numpy as np

In [ ]: model = load_model("/content/flowers.h1")

In [ ]: #Testing with a random rose image from Google
        img = image.load_img("/content/rose.gif", target_size = (64,64) )

In [ ]: img
Out[ ]: 
```

```
In [ ]: x = image.img_to_array(img)
        x.ndim
Out[ ]: 3

In [ ]: x = np.expand_dims(x,axis = 0)
        x.ndim
Out[ ]: 4

In [ ]: pred = model.predict(x)
        pred
Out[ ]: array([[0., 0., 1., 0., 0.]], dtype=float32)

In [ ]: labels = ['daisy','dandelion','roses','sunflowers','tulips']

In [ ]: labels[np.argmax(pred)]
Out[ ]: 'roses'
```