

# ASSIGNMENT 3

ASSIGNMENT DATE	01 OCTOBER 2020
STUDENT NAME	SHAKEEL MOHAMMED G
STUDENT ROLL NUMBER	2019504584
MAXIMUM MARKS	2 MARKS

Problem Statement:- Build CNN Model for Classification Of Flowers

Perform Below Tasks to complete the assignment: -

1. Download the Dataset
2. Image Augmentation
3. Create Model
4. Add Layers (Convolution, MaxPooling, Flatten,Dense-(Hidden Layers), Output)
5. Compile The Model
6. Fit The Model
7. Save The Model
8. Test The Model

```
In [ ]: from google.colab import drive
drive.mount('/content/drive')
```

Mounted at /content/drive

```
In [ ]: !pip install split_folders
import splitfolders

input_folder = "/content/drive/MyDrive/IBM/Flowers" #Enter Input Folder
output = "/content/drive/MyDrive/IBM/Flowers/data" #Enter Output Folder

splitfolders.ratio(input_folder, output=output, seed=42, ratio=(0.8,0.2))
```

Looking in indexes: <https://pypi.org/simple>, <https://us-python.pkg.dev/colab-wheels/public/simple/>

Requirement already satisfied: split\_folders in /usr/local/lib/python3.7/dist-packages (0.5.1)

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## Importing the libraries

```
In [ ]: from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense
```

```
In [ ]: from tensorflow.keras.layers import Convolution2D
from tensorflow.keras.layers import MaxPooling2D
from tensorflow.keras.layers import Flatten
```

```
In [ ]: #import the preprocess library of image
        from tensorflow.keras.preprocessing.image import ImageDataGenerator
```

## Image Augmentation

```
In [ ]: train_datagen = ImageDataGenerator(rescale=1./255, shear_range=0.2, zoom_range=0.2, horizontal_flip=True)
        #rescale = pixel value rescaling to 0 to 1 from 0 to 255
        #shear_range => counter clock wise rotation(anti clock)
```

```
In [ ]: test_datagen = ImageDataGenerator(rescale=1./255)
```

```
In [ ]: #Load your images data
```

```
In [ ]: #Load your images data
```

```
In [ ]: x_train = train_datagen.flow_from_directory('/content/drive/MyDrive/IBM/Flowers/data/train')
        Found 3452 images belonging to 5 classes.
```

```
In [ ]: x_test = test_datagen.flow_from_directory('/content/drive/MyDrive/IBM/Flowers/data/test')
        Found 865 images belonging to 5 classes.
```

```
In [ ]: x_train.class_indices
```

```
Out[ ]: {'daisy': 0, 'dandelion': 1, 'rose': 2, 'sunflower': 3, 'tulip': 4}
```

## Create Model

```
In [ ]: #initialize the model
        model = Sequential()
```

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

```
In [ ]: #add convolution layer
        model.add(Convolution2D(32, (3,3), input_shape=(128,128,3), activation='relu'))
        # 32 => no of feature detectors
        #(3,3)=> kernel size(feature detector size => 3*3 matrix)
```

```
In [ ]: #add maxpooling layer
        model.add(MaxPooling2D(pool_size=(2,2)))
```

```
In [ ]: # you can add more convolutional and pooling layers
        model.add(Convolution2D(32, (3,3), input_shape=(128,128,3), activation='relu'))
        model.add(MaxPooling2D(pool_size=(2,2)))
```

```
In [ ]: #flatten layer => input layer to your ANN
        model.add(Flatten())
```

```
In [ ]: #hidden layers
model.add(Dense(units=500,kernel_initializer="random_uniform",activation="relu"))
model.add(Dense(units=200,kernel_initializer="random_uniform",activation="relu"))
model.add(Dense(units=300,kernel_initializer="random_uniform",activation="relu"))
model.add(Dense(units=400,kernel_initializer="random_uniform",activation="relu"))

In [ ]: #output layer
model.add(Dense(units=5,kernel_initializer="random_uniform",activation="softmax"))
```

## Compile The Model

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

## Fit The Model

```
In [ ]: #train the model
model.fit(x_train,steps_per_epoch=len(x_train),epochs=10,validation_data=x_test,validation_steps=len(x_test))
#steps_per_epoch = no of train images/batch size
#validation_steps = no of test images/batch size
```

```
Epoch 1/10
35/35 [=====] - 106s 3s/step - loss: 1.4203 - accuracy: 0.3543 - val_loss: 1.1715 - val_accuracy: 0.4775
Epoch 2/10
35/35 [=====] - 99s 3s/step - loss: 1.2220 - accuracy: 0.4615 - val_loss: 1.1513 - val_accuracy: 0.4902
Epoch 3/10
35/35 [=====] - 100s 3s/step - loss: 1.1233 - accuracy: 0.5159 - val_loss: 1.0047 - val_accuracy: 0.5653
Epoch 4/10
35/35 [=====] - 98s 3s/step - loss: 1.0440 - accuracy: 0.5744 - val_loss: 0.9395 - val_accuracy: 0.6266
Epoch 5/10
35/35 [=====] - 100s 3s/step - loss: 1.0361 - accuracy: 0.5759 - val_loss: 0.9530 - val_accuracy: 0.6162
Epoch 6/10
35/35 [=====] - 99s 3s/step - loss: 1.0050 - accuracy: 0.6037 - val_loss: 0.8916 - val_accuracy: 0.6439
Epoch 7/10
35/35 [=====] - 101s 3s/step - loss: 0.9490 - accuracy: 0.6295 - val_loss: 0.8843 - val_accuracy: 0.6416
Epoch 8/10
35/35 [=====] - 98s 3s/step - loss: 0.8924 - accuracy: 0.6550 - val_loss: 0.8537 - val_accuracy: 0.6590
Epoch 9/10
35/35 [=====] - 99s 3s/step - loss: 0.8786 - accuracy: 0.6602 - val_loss: 0.8756 - val_accuracy: 0.6451
Epoch 10/10
35/35 [=====] - 101s 3s/step - loss: 0.8491 - accuracy: 0.6790 - val_loss: 0.8410 - val_accuracy: 0.6486
```

```
Out[ ]: <keras.callbacks.History at 0x7fc4e65a0310>
```

# Save The Model

```
In [ ]: model.save("/flowers.h5")
```

# 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("/flowers.h5")
```

```
In [ ]: img = image.load_img("/content/drive/MyDrive/IBM/sunflower.jpg",target_size=(128,128))
```

```
In [ ]: img
```

```
Out[ ]:
```



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

```
In [ ]: x
```

```

Out[ ]: array([[ 94., 127.,  56.],
               [ 92., 125.,  54.],
               [ 90., 123.,  52.],
               ...,
               [ 96., 128.,  52.],
               [104., 135.,  59.],
               [112., 140.,  65.]],

               [[106., 133.,  64.],
               [109., 136.,  67.],
               [109., 136.,  67.],
               ...,
               [101., 132.,  54.],
               [111., 139.,  62.],
               [115., 142.,  65.]],

               [[129., 150.,  85.],
               [130., 151.,  86.],
               [132., 153.,  88.],
               ...,
               [108., 137.,  53.],
               [112., 141.,  59.],
               [120., 144.,  66.]],

               ...,

               [[141., 159., 111.],
               [134., 153.,  98.],
               [125., 145.,  86.],
               ...,
               [ 62.,  96.,   2.],
               [ 55.,  88.,   7.],
               [ 48.,  82.,   8.]],

               [[141., 158., 113.],
               [138., 155., 110.],
               [132., 150., 102.],
               ...,
               [ 62.,  96.,   2.],
               [ 55.,  88.,   7.],
               [ 47.,  81.,   7.]],

               [[133., 152., 106.],
               [128., 150., 101.],
               [116., 140.,  88.],
               ...,
               [ 61.,  94.,   3.],
               [ 57.,  89.,   6.],
               [ 50.,  80.,  10.]])], dtype=float32)

```

```
In [ ]: x.shape
```

```
Out[ ]: (128, 128, 3)
```

```
In [ ]: #(1,64,64,3) to expand the dims
```

```
In [ ]: x = np.expand_dims(x,axis=0)
        x.shape
```

```
Out[ ]: (1, 128, 128, 3)
```

```
In [ ]: pred_prob = model.predict(x)
1/1 [=====] - 0s 168ms/step
```

```
In [ ]: pred_prob
```

```
Out[ ]: array([[0., 0., 0., 1., 0.]], dtype=float32)
```

```
In [ ]: class_name=["DAISY","DANDELION",'ROSE','SUNFLOWER',"TULIP",]
pred_id = pred_prob.argmax(axis=1)[0]
```

```
In [ ]: pred_id
```

```
Out[ ]: 3
```

```
In [ ]: print("Predicted Flower is",str(class_name[pred_id]))
Predicted Flower is SUNFLOWER
```