#Assignment 3

- Download the Dataset: https://drive.google.com/file/d/1xkynpL15pt6KT3YSlDimu4A5iRU9qYck/view
- Image Augmentation
- Create Model
- Add Layers (Convolution, MaxPooling, Flatten, Dense-(Hidden Layers), Output)
- Compile The Model
- Fit The Model
- Save The Model
- Test The Model

```
#Importing Packages
```

```
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import
Convolution2D, MaxPooling2D, Flatten, Dense
from tensorflow.keras.preprocessing.image import ImageDataGenerator as
idm
import numpy as np
import warnings
#Supressing warnings
warnings.filterwarnings('ignore')
```

2.Image Augmentation

#3.Create Model

```
# Creating augmentation on training variable
train_flowers=idm(rescale=1./255,zoom_range=0.2,horizontal_flip=True)

# Passing training data to train variable
Xtrain =
train_flowers.flow_from_directory('/content/drive/MyDrive/IBM/Flowers-Dataset',target_size=(76,76),class_mode='categorical',batch_size=100)
Found 4141 images belonging to 5 classes.

# Creating augmentation on testing variable
test_flowers=idm(rescale=1./255)

# Passing testing data to test variable
Xtest =
test_flowers.flow_from_directory('/content/drive/MyDrive/IBM/Flower_Training',target_size=(76,76),class_mode='categorical',batch_size=100)
Found 204 images belonging to 5 classes.
```

```
Flower model = Sequential()
Flower model.add(Convolution2D(32,
(3,3),activation='relu',input_shape=(76,76,3)))
Flower model.add(MaxPooling2D(pool size=(2,2)))
Flower model.add(Flatten())
Flower model.add(Dense(300,activation='relu'))
Flower model.add(Dense(150,activation='relu'))
Flower model.add(Dense(5,activation='softmax'))
#4. Compile the Model
Flower_model.compile(optimizer='adam',loss='categorical crossentropy',
metrics=['accuracy'])
#5. Fit the Model
Flower model.fit generator(Xtrain, steps per epoch= len
(Xtrain),epochs= 10,validation_data=Xtest,validation_steps= len
(Xtest))
Epoch 1/10
- accuracy: 0.3700 - val loss: 1.1356 - val accuracy: 0.5490
Epoch 2/10
- accuracy: 0.5412 - val loss: 1.1446 - val accuracy: 0.6422
Epoch 3/10
- accuracy: 0.6042 - val loss: 1.1835 - val accuracy: 0.6225
Epoch 4/10
- accuracy: 0.6264 - val loss: 1.0033 - val accuracy: 0.6765
Epoch 5/10
- accuracy: 0.6619 - val loss: 0.9993 - val accuracy: 0.7059
Epoch 6/10
- accuracy: 0.6783 - val loss: 0.9690 - val accuracy: 0.7206
Epoch 7/10
- accuracy: 0.6923 - val loss: 0.8731 - val accuracy: 0.7059
Epoch 8/10
- accuracy: 0.7073 - val loss: 1.0149 - val accuracy: 0.6667
Epoch 9/10
- accuracy: 0.7242 - val_loss: 0.9583 - val_accuracy: 0.6863
Epoch 10/10
- accuracy: 0.7262 - val loss: 0.9150 - val accuracy: 0.7206
```

```
<keras.callbacks.History at 0x7fd5aec82f50>
#7. Save the model
Flower model.save('Flower.h5')
#8. Test the model
test img=image.load img('/content/drive/MyDrive/IBM/Flowers-Dataset/
sunflower/200557977 bf24d9550b.jpg',target size=(76,76))
test imq
x=image.img to array(test_img)
x=np.expand dims(x,axis=0)
predicted=np.argmax(Flower model.predict(x))
Prediction category=['daisy','dandelion','rose','sunflower','tulip']
Prediction category[predicted]
{"type":"string"}
test img1=image.load img('/content/drive/MyDrive/IBM/Flowers-Dataset/
daisy/1140299375 3aa7024466.jpg',target size=(76,76))
test img1
x=image.img_to_array(test_img1)
x=np.expand dims(x,axis=0)
predicted=np.argmax(Flower model.predict(x))
Prediction category[predicted]
{"type": "string"}
test_img2=image.load_img('/content/drive/MyDrive/IBM/Flowers-Dataset/
rose/7251352826 69b62cba2c m.jpg',target size=(76,76))
test img2
```



```
x=image.img_to_array(test_img2)
x=np.expand_dims(x,axis=0)
predicted=np.argmax(Flower_model.predict(x))
Prediction_category[predicted]
{"type":"string"}
```