

Assignment –3

Build CNN for Classification of Flowers

Import required packages

```
[ ] import keras
```

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

Mounted at /content/drive

```
[ ] from keras.preprocessing.image import ImageDataGenerator
```

```
[ ] train_datagen=ImageDataGenerator(rescale=1./255,  
                                     shear_range=0.2,  
                                     rotation_range=180,  
                                     zoom_range=0.2,  
                                     horizontal_flip=True)
```

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

```
[ ] #install Kaggle  
!pip install -q kaggle
```

```
[ ] #create a kaggle folder  
!mkdir ~/.kaggle
```

```
[ ] #copy the kaggle.json to folder created  
!cp kaggle.json ~/.kaggle/
```

cp: cannot stat 'kaggle.json': No such file or directory

```
[ ] #permission for the json to act  
! chmod 600 ~/.kaggle/kaggle.json
```

chmod: cannot access '/root/.kaggle/kaggle.json': No such file or directory

1.Download the dataset

```
▶ x_train = train_datagen.flow_from_directory(r'/content/drive/MyDrive/IBM/train_set',
target_size = (128,128),
batch_size = 32,
class_mode = 'binary')
```

Found 2313 images belonging to 5 classes.

```
[ ] x_test = test_datagen.flow_from_directory(r'/content/drive/MyDrive/IBM/test_set',
target_size = (128,128),
batch_size = 32,
class_mode = 'binary')
```

Found 2068 images belonging to 5 classes.

2. Image Augmentation

```
[ ] #give any random image path

img = image.load_img(r'/content/drive/MyDrive/IBM/test_set/daisy/10300722094_28fa978807_n.jpg')

x = image.img_to_array(img)

#expand the image shape

x = np.expand_dims(x,axis= 0)
```

```
[ ] img
```



```
[ ] pip install imgaug
```

```
Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simple/
Requirement already satisfied: imgaug in /usr/local/lib/python3.7/dist-packages (0.4.0)
Requirement already satisfied: scipy in /usr/local/lib/python3.7/dist-packages (from imgaug) (1.7.3)
Requirement already satisfied: scikit-image>=0.14.2 in /usr/local/lib/python3.7/dist-packages (from imgaug) (0.18.3)
Requirement already satisfied: matplotlib in /usr/local/lib/python3.7/dist-packages (from imgaug) (3.2.2)
Requirement already satisfied: Shapely in /usr/local/lib/python3.7/dist-packages (from imgaug) (1.8.4)
Requirement already satisfied: imageio in /usr/local/lib/python3.7/dist-packages (from imgaug) (2.9.0)
Requirement already satisfied: six in /usr/local/lib/python3.7/dist-packages (from imgaug) (1.15.0)
Requirement already satisfied: opencv-python in /usr/local/lib/python3.7/dist-packages (from imgaug) (4.6.0.66)
Requirement already satisfied: numpy>=1.15 in /usr/local/lib/python3.7/dist-packages (from imgaug) (1.21.6)
Requirement already satisfied: Pillow in /usr/local/lib/python3.7/dist-packages (from imgaug) (7.1.2)
Requirement already satisfied: PyWavelets>=1.1.1 in /usr/local/lib/python3.7/dist-packages (from scikit-image>=0.14.2->imgaug) (1.3.0)
Requirement already satisfied: networkx>=2.0 in /usr/local/lib/python3.7/dist-packages (from scikit-image>=0.14.2->imgaug) (2.6.3)
Requirement already satisfied: kiwisolver>=1.0.1 in /usr/local/lib/python3.7/dist-packages (from matplotlib->imgaug) (1.4.4)
Requirement already satisfied: pyparsing<2.0.4,!=2.1.2,!=2.1.6,>=2.0.1 in /usr/local/lib/python3.7/dist-packages (from matplotlib->imgaug) (3.0.9)
Requirement already satisfied: python-dateutil>=2.1 in /usr/local/lib/python3.7/dist-packages (from matplotlib->imgaug) (2.8.2)
Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.7/dist-packages (from matplotlib->imgaug) (0.11.0)
Requirement already satisfied: typing-extensions in /usr/local/lib/python3.7/dist-packages (from kiwisolver>=1.0.1->matplotlib->imgaug) (4.1.1)
```

```
[ ] pip install ipyplot
```

```
Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simple/
Collecting ipyplot
  Downloading ipyplot-1.1.1-py3-none-any.whl (13 kB)
Requirement already satisfied: pillow in /usr/local/lib/python3.7/dist-packages (from ipyplot) (7.1.2)
Requirement already satisfied: IPython in /usr/local/lib/python3.7/dist-packages (from ipyplot) (7.9.0)
Collecting shortuid
  Downloading shortuid-1.0.9-py3-none-any.whl (9.4 kB)
Requirement already satisfied: numpy in /usr/local/lib/python3.7/dist-packages (from ipyplot) (1.21.6)
Requirement already satisfied: backcall in /usr/local/lib/python3.7/dist-packages (from IPython->ipyplot) (0.2.0)
Requirement already satisfied: decorator in /usr/local/lib/python3.7/dist-packages (from IPython->ipyplot) (4.4.2)
Requirement already satisfied: pexpect in /usr/local/lib/python3.7/dist-packages (from IPython->ipyplot) (4.8.0)
Requirement already satisfied: pickleshare in /usr/local/lib/python3.7/dist-packages (from IPython->ipyplot) (0.7.5)
Requirement already satisfied: traitlets>=4.2 in /usr/local/lib/python3.7/dist-packages (from IPython->ipyplot) (5.1.1)
Requirement already satisfied: setuptools>=18.5 in /usr/local/lib/python3.7/dist-packages (from IPython->ipyplot) (57.4.0)
Requirement already satisfied: prompt-toolkit<2.1.0,>=2.0.0 in /usr/local/lib/python3.7/dist-packages (from IPython->ipyplot) (2.0.10)
Requirement already satisfied: pygments in /usr/local/lib/python3.7/dist-packages (from IPython->ipyplot) (2.6.1)
Collecting jedi>=0.10
  Downloading jedi-0.18.1-py2.py3-none-any.whl (1.6 MB)
Requirement already satisfied: parso<0.9.0,>=0.8.0 in /usr/local/lib/python3.7/dist-packages (from jedi>=0.10->IPython->ipyplot) (0.8.3)
Requirement already satisfied: six>=1.9.0 in /usr/local/lib/python3.7/dist-packages (from prompt-toolkit<2.1.0,>=2.0.0->IPython->ipyplot) (1.15.0)
Requirement already satisfied: wcwidth in /usr/local/lib/python3.7/dist-packages (from prompt-toolkit<2.1.0,>=2.0.0->IPython->ipyplot) (0.2.5)
Requirement already satisfied: ptyprocess>=0.5 in /usr/local/lib/python3.7/dist-packages (from pexpect->IPython->ipyplot) (0.7.0)
Installing collected packages: jedi, shortuid, ipyplot
Successfully installed ipyplot-1.1.1 jedi-0.18.1 shortuid-1.0.9
```

3.Create Model

```
[ ] import ipyplot
import imageio
import imgaug as ia
import imgaug.augmenters as iaa
```

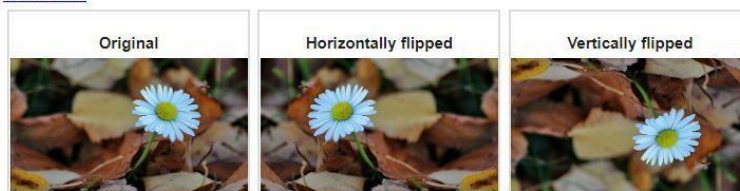
WARNING! Google Colab Environment detected!
You might encounter issues while running in Google Colab environment.
If images are not displaying properly please try setting `force_b64` param to `True`.

```
[ ] input=imageio.imread("/content/drive/MyDrive/IBM/test_set/daisy/10559679065_50d2b16f6d.jpg")
```

```
[ ] hflip = iaa.Fliplr(p=1.0)
input_hf = hflip.augment_image(input)
```

```
[ ] vflip=iaa.Flipud (p=1.0)
input_vf=vflip.augment_image(input)
images_list=[input, input_hf, input_vf]
labels=['Original', 'Horizontally flipped', 'Vertically flipped']
ipyplot.plot_images (images_list,labels=labels, img_width=180)
```

[show html](#)



```
[ ] noise=iaa.AdditiveGaussianNoise (18,48)
input_noise=noise.augment_image(input)
images_list=[input, input_noise]
labels= ["Original", "Gaussian Noise Image"]
ipyplot.plot_images(images_list, labels=labels, img_width=188)
```

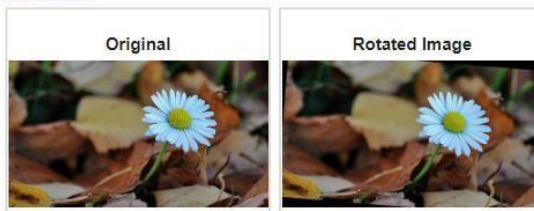
[show html](#)



```
[ ] rot1 = iaa.Affine(rotate=(-30,30))

input_rot1 = rot1.augment_image(input)
images_list=[input, input_rot1]
labels= ['Original', 'Rotated Image']
ipyplot.plot_images(images_list,labels=labels, img_width=180)
```

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

```
[ ] #To define Linear intialisation import

from keras.models import Sequential
#To add Layers import Dense

from keras.layers import Dense

#To create Convolution kernel import Convolution2D from keras.layers import Convolution2D

from keras.layers import Convolution2D

from keras.layers import MaxPooling2D

#import Flatten Layer

from keras.layers import Flatten

import warnings
warnings.filterwarnings('ignore')
```

```
[ ] #initialize our model
model = Sequential()
```

```
[ ] #Adding Convolutional Layer
model.add(Convolution2D(32,(3,3),input_shape=(128,128,3),activation='relu'))
```

```
[ ] #Adding Pooling LayerMax Pooling
model.add(MaxPooling2D(pool_size=(2,2)))
```

```
[ ] #Adding Flatten Layer
model.add(Flatten())
```

Adding Hidden Layers

```
[ ] model.add(Dense(2,activation='relu'))
```

```
[ ] model.add(Dense(150, bias_initializer='uniform', activation='relu'))
```

Adding Output Layer

```
[ ] model.add(Dense (1, bias_initializer='uniform', activation='sigmoid'))
```

5.Compile the Model

```
[ ]
model.compile(loss = 'binary_crossentropy',

              optimizer = "adam",

              metrics = ["accuracy"])
```

6. Fit the Model

```
[ ] model.fit_generator(x_train, steps_per_epoch=14,

                      epochs=10, validation_data=x_test,

                      validation_steps=4)

Epoch 1/10
14/14 [=====] - 107s 8s/step - loss: -23.7952 - accuracy: 0.2232 - val_loss: -77.4922 - val_accuracy: 0.2578
Epoch 2/10
14/14 [=====] - 84s 6s/step - loss: -235.4541 - accuracy: 0.2282 - val_loss: -615.0518 - val_accuracy: 0.2109
Epoch 3/10
14/14 [=====] - 75s 6s/step - loss: -1322.1671 - accuracy: 0.2500 - val_loss: -2788.8315 - val_accuracy: 0.1875
Epoch 4/10
14/14 [=====] - 62s 5s/step - loss: -4095.4810 - accuracy: 0.2282 - val_loss: -7051.1777 - val_accuracy: 0.2266
Epoch 5/10
14/14 [=====] - 53s 4s/step - loss: -13293.1729 - accuracy: 0.2121 - val_loss: -19571.5312 - val_accuracy: 0.2812
Epoch 6/10
14/14 [=====] - 45s 3s/step - loss: -21887.4746 - accuracy: 0.2723 - val_loss: -33551.9688 - val_accuracy: 0.2891
Epoch 7/10
14/14 [=====] - 47s 4s/step - loss: -49263.8320 - accuracy: 0.2277 - val_loss: -76996.9688 - val_accuracy: 0.2734
Epoch 8/10
14/14 [=====] - 41s 3s/step - loss: -96499.0391 - accuracy: 0.2478 - val_loss: -121640.2500 - val_accuracy: 0.2266
Epoch 9/10
14/14 [=====] - 31s 2s/step - loss: -181292.6250 - accuracy: 0.2411 - val_loss: -259279.2188 - val_accuracy: 0.2422
Epoch 10/10
14/14 [=====] - 28s 2s/step - loss: -281458.2188 - accuracy: 0.2344 - val_loss: -333128.7812 - val_accuracy: 0.2812
<keras.callbacks.History at 0x7f52a5860e00>
```

7. Save the Model & Test the Model

```
[ ] model.save("flowers.h5")
```

```
[ ] ls
```

```
drive/ flowers.h5 sample_data/
```

```
[ ] from keras.models import load_model

#import image class from keros

from keras.preprocessing import image

#import numpy

import numpy as np

#import cv2

import cv2
```

```
[ ] #Load the saved model

model = load_model("flowers.h5")
```