

Assignment 3

Build CNN Model for Classification of Flowers

1. Download the Dataset

```
pwd
In [41]:

'/content/drive/MyDrive'
Out[41]:

Load the Image Dataset

ls
In [ ]:

drive/  sample_data/
In [ ]:

from google.colab import drive
drive.mount('/content/drive')
Mounted at /content/drive
Un-zip the Folder

cd /content/drive/MyDrive
In [ ]:

/content/drive/MyDrive
In [77]:

!unzip Flowers-Dataset.zip
Archive:  Flowers-Dataset.zip
replace flowers/daisy/100080576_f52e8ee070_n.jpg? [y]es, [n]o, [A]ll, [N]on
e, [r]ename: N
In [ ]:

pwd
Out[ ]:

'/content/drive/MyDrive'
```

2. Image Augmentation

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

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

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

pwd
Out[ ]:

'/content/drive/MyDrive'
In [ ]:
```

```
x_train=train_datagen.flow_from_directory(r"/content/drive/MyDrive/flowers"
,target_size=(64,64),class_mode='categorical',batch_size=24)
```

Found 4317 images belonging to 5 classes.

In []:

```
x_test=test_datagen.flow_from_directory(r"/content/drive/MyDrive/flowers",
target_size=(64,64),class_mode='categorical',batch_size=24)
```

Found 4317 images belonging to 5 classes.

In []:

```
x_train.class_indices
```

Out[]:

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

CNN

3. Create Model

In []:

```
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import
Dense,Convolution2D,MaxPooling2D,Flatten,Dense
```

In []:

```
model=Sequential()
```

4. Add Layers(Convolution, MaxPooling, Flatten)

In []:

```
model.add(Convolution2D(32,(3,3),input_shape=(64,64,3),activation='relu'))
```

In []:

```
model.add(MaxPooling2D(pool_size=(2,2)))
```

In []:

```
model.add(Flatten())
```

In []:

```
model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
conv2d (Conv2D)	(None, 62, 62, 32)	896
max_pooling2d (MaxPooling2D)	(None, 31, 31, 32)	0
flatten (Flatten)	(None, 30752)	0
=====		
Total params: 896		
Trainable params: 896		
Non-trainable params: 0		
=====		

In []:

```
32*(3*3*3+1)
```

896

Out[]:

Dense - (Hidden Layers)

In []:

```
model.add(Dense(300,activation='relu'))
model.add(Dense(150,activation='relu'))
```

Output Layers

In []:

```
model.add(Dense(5,activation='softmax'))
```

5. Compile the model

In []:

```
model.compile(loss='categorical_crossentropy',metrics=['accuracy'],optimizer='adam')
```

In []:

```
len(x_train)
```

Out[]:

180

In []:

4317/24

Out[]:

179.875

6. Fit the Model

In []:

```
model.fit(x_train, epochs = 5, validation_data=x_test,
steps_per_epoch=len(x_train), validation_steps=len(x_test))
```

Epoch 1/5

180/180 [=====] - 711s 4s/step - loss: 1.6647 - accuracy: 0.2201 - val_loss: 1.6395 - val_accuracy: 0.2437

Epoch 2/5

180/180 [=====] - 65s 362ms/step - loss: 1.6257 - accuracy: 0.2409 - val_loss: 1.6142 - val_accuracy: 0.2437

Epoch 3/5

180/180 [=====] - 66s 366ms/step - loss: 1.6083 - accuracy: 0.2437 - val_loss: 1.6034 - val_accuracy: 0.2437

Epoch 4/5

180/180 [=====] - 65s 361ms/step - loss: 1.6015 - accuracy: 0.2437 - val_loss: 1.5998 - val_accuracy: 0.2437

Epoch 5/5

180/180 [=====] - 65s 360ms/step - loss: 1.5994 - accuracy: 0.2432 - val_loss: 1.5987 - val_accuracy: 0.2437

Out[]:

<keras.callbacks.History at 0x7fb054985e90>

7. Save the Model

```
model.save('flowers.h5')
```

In [39]:

```
ls flowers/  
daisy/  dandelion/  rose/  sunflower/  tulip/
```

In [40]:

8. Test the Model

```
import numpy as np  
from tensorflow.keras.models import load_model  
from tensorflow.keras.preprocessing import image
```

In [42]:

```
#load the model  
model=load_model('flowers.h5')
```

In [43]:

```
img=image.load_img(r"/content/drive/MyDrive/flowers/daisy/100080576_f52e8ee  
070_n.jpg")
```

In [44]:

```
img
```

In [45]:

Out[45]:



```
img=image.load_img(r"/content/drive/MyDrive/flowers/daisy/100080576_f52e8ee  
070_n.jpg", target_size=(64,64))  
img
```

In [46]:

Out[46]:



```
x=image.img_to_array(img)
```

In [47]:

In [48]:

x

Out[48]:

```
array([[141., 141., 139.],
       [149., 149., 149.],
       [152., 152., 154.],
       ...,
       [162., 161., 166.],
       [154., 154., 152.],
       [153., 153., 153.]],

      [[136., 135., 131.],
       [146., 145., 143.],
       [169., 168., 174.],
       ...,
       [159., 158., 163.],
       [155., 155., 153.],
       [149., 149., 149.]],

      [[125., 125., 117.],
       [138., 140., 137.],
       [152., 152., 152.],
       ...,
       [156., 156., 156.],
       [157., 157., 155.],
       [143., 142., 140.]],

      ...,

      [[ 41.,  44.,  23.],
       [ 43.,  46.,  25.],
       [ 49.,  51.,  37.],
       ...,
       [128., 124., 121.],
       [125., 121., 118.],
       [125., 122., 117.]],

      [[ 43.,  46.,  25.],
       [ 43.,  46.,  25.],
       [ 54.,  55.,  37.],
       ...,
       [130., 126., 125.],
       [129., 125., 124.],
       [127., 123., 122.]],

      [[ 44.,  47.,  26.],
       [ 45.,  48.,  27.],
       [ 53.,  55.,  34.],
       ...,
       [137., 133., 132.],
       [133., 129., 128.],
       [130., 126., 125.]])], dtype=float32)
```

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

In [49]:

In [50]:

x

Out[50]:

```
array([[[[141., 141., 139.],
         [149., 149., 149.],
         [152., 152., 154.],
         ...,
         [162., 161., 166.],
         [154., 154., 152.],
         [153., 153., 153.]],

        [[136., 135., 131.],
         [146., 145., 143.],
         [169., 168., 174.],
         ...,
         [159., 158., 163.],
         [155., 155., 153.],
         [149., 149., 149.]],

        [[125., 125., 117.],
         [138., 140., 137.],
         [152., 152., 152.],
         ...,
         [156., 156., 156.],
         [157., 157., 155.],
         [143., 142., 140.]],

        ...,

        [[ 41.,  44.,  23.],
         [ 43.,  46.,  25.],
         [ 49.,  51.,  37.],
         ...,
         [128., 124., 121.],
         [125., 121., 118.],
         [125., 122., 117.]],

        [[ 43.,  46.,  25.],
         [ 43.,  46.,  25.],
         [ 54.,  55.,  37.],
         ...,
         [130., 126., 125.],
         [129., 125., 124.],
         [127., 123., 122.]],

        [[ 44.,  47.,  26.],
         [ 45.,  48.,  27.],
         [ 53.,  55.,  34.],
         ...,
         [137., 133., 132.],
         [133., 129., 128.],
         [130., 126., 125.]]]], dtype=float32)
```

```
y=np.argmax(model.predict(x),axis=0)
```

In [70]:

In [52]:

y

```
array([1])
```

Out[52]:

```
x_train.class_indices
```

In [53]:

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

Out[53]:

```
index=['daisy','dandelion','rose','sunflower']
```

In [54]:

```
index[y[0]]
```

In [71]:

```
'daisy'
```

Out[71]:

```
img=image.load_img(r"/content/drive/MyDrive/flowers/dandelion/10200780773_c
6051a7d71_n.jpg", target_size=(64,64))
x=image.img_to_array(img)
x=np.expand_dims(x,axis=0)
y=np.argmax(model.predict(x),axis=1)
index=['daisy','dandelion','rose','sunflower']
index[y[0]]
```

In [61]:

```
'dandelion'
```

Out[61]:

```
img
```

In [57]:



Out[57]:

```
img=image.load_img(r"/content/drive/MyDrive/flowers/rose/10503217854_e66a80
4309.jpg", target_size=(64,64))
x=image.img_to_array(img)
x=np.expand_dims(x,axis=0)
y=np.argmax(model.predict(x),axis=1)
index=['daisy','rose','dandelion','sunflower']
index[y[0]]
```

In [74]:

```
'rose'
```

Out[74]:

```
img
```

In [75]:



Out[75]:

```
img=image.load_img(r"/content/drive/MyDrive/flowers/sunflower/10386503264_e
05387e1f7_m.jpg", target_size=(64,64))
x=image.img_to_array(img)
x=np.expand_dims(x,axis=0)
y=np.argmax(model.predict(x),axis=0)
```

In [72]:

```
index=['sunflower','daisy','dandelion','rose']  
index[y[0]]
```

'sunflower'

img



Out[72]:

In [60]:

Out[60]: