

## Assignment 3

### Build CNN Model for Classification of Flowers

#### 1. Download the Dataset

```
pwd
```

```
{"type": "string"}
```

##### Load the Image Dataset

```
ls
```

```
drive/ sample_data/
```

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

```
Mounted at /content/drive
```

##### Un-zip the Folder

```
cd /content/drive/MyDrive
```

```
/content/drive/MyDrive
```

```
!unzip Flowers-Dataset.zip
```

```
Archive:  Flowers-Dataset.zip  
replace flowers/daisy/100080576_f52e8ee070_n.jpg? [y]es, [n]o, [A]ll,  
[N]one, [r]ename: N
```

```
pwd
```

```
{"type": "string"}
```

#### 2. Image Augmentation

```
from tensorflow.keras.preprocessing.image import ImageDataGenerator
```

```
train_datagen=ImageDataGenerator(rescale=1./255,zoom_range=0.2,horizon  
tal_flip=True,vertical_flip=False)
```

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

```
pwd
```

```
{"type": "string"}
```

```
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.
```

```
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.

```
x_train.class_indices
```

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

## CNN

### 3. Create Model

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

### 4. Add Layers(Convolution, MaxPooling, Flatten)

```
model.add(Convolution2D(32,
(3,3),input_shape=(64,64,3),activation='relu'))
model.add(MaxPooling2D(pool_size=(2,2)))
model.add(Flatten())
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		

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

896

### Dense - (Hidden Layers)

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

## Output Layers

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

## 5. Compile the model

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

```
len(x_train)
```

```
180
```

```
4317/24
```

```
179.875
```

## 6. Fit the Model

```
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
```

```
<keras.callbacks.History at 0x7fb054985e90>
```

## 7. Save the Model

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

```
ls flowers/
```

```
daisy/  dandelion/  rose/  sunflower/  tulip/
```

## 8. Test the Model

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

```
#load the model
```

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

```
img=image.load_img(r"/content/drive/MyDrive/flowers/daisy/100080576_f52e8ee070_n.jpg")
```

img



```
img=image.load_img(r"/content/drive/MyDrive/flowers/daisy/100080576_f52e8ee070_n.jpg", target_size=(64,64))
```

img



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

x

```
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)

```

```

x

```

```

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)

y
array([1])

x_train.class_indices

```

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

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

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

```
{"type":"string"}
```

```
img=image.load_img(r"/content/drive/MyDrive/flowers/dandelion/  
10200780773_c6051a7d71_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]]
```

```
{"type":"string"}
```

```
img
```



```
img=image.load_img(r"/content/drive/MyDrive/flowers/rose/  
10503217854_e66a804309.jpg", target_size=(64,64))
```

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

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

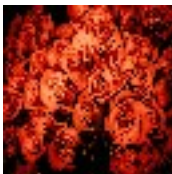
```
y=np.argmax(model.predict(x),axis=1)
```

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

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

```
{"type":"string"}
```

```
img
```



```
img=image.load_img(r"/content/drive/MyDrive/flowers/sunflower/  
10386503264_e05387elf7_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)
```

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

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

```
{"type":"string"}
```

img

