Assignment 3

Build CNN Model for Classification of Flowers

DATE	**9 october 2022**
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PROJECT NAME	**Fertilizer recommendation system for disease prediction**
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1. Download the Dataset

In [41]:

```
pwd
```

Out[41]:

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

Load the Image Dataset

In [ ]:

```
ls
```

```
drive/  sample_data/
```

In [ ]:

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

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

Un-zip the Folder

In [ ]:

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

```
/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

In [ ]:
```
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)
```

Out[ ]:

```
896
```

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'],optimize
r='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 - ac
curacy: 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

In [39]:

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

In [40]:

```
ls flowers/
```

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

8. Test the Model

In [42]:

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

In [43]:

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

In [44]:

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

In [45]:

```
img
```

Out[45]:

In [46]:

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

Out[46]:

In [47]:

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

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)
```

In [49]:

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

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)
```

In [70]:

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

In [52]:

```
y
```

Out[52]:

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

In [53]:

```
x_train.class_indices
```

Out[53]:

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

In [54]:

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

In [71]:

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

Out[71]:

```
'daisy'
```

In [61]:

```
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]]
```

Out[61]:

```
'dandelion'
```

In [57]:

```
img
```

Out[57]:

In [74]:

```
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]]
```

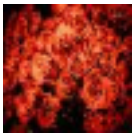
Out[74]:

```
'rose'
```

In [75]:

```
img
```

Out[75]:



In [72]:

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

Out[72]:

```
'sunflower'
```

In [60]:

```
img
```

Out[60]: