import the libraries

```
In [15]:
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import
Dense, Convolution 2D, Max Pooling 2D, Flatten
import numpy as np
from tensorflow.keras.models import load model
from tensorflow.keras.preprocessing import image
Image Augmentation
                                                                       In [3]:
from tensorflow.keras.preprocessing.image import ImageDataGenerator
train datagen=ImageDataGenerator(rescale=1./255,zoom range=0.2,horizontal f
lip=True, vertical flip=False)
test datagen=ImageDataGenerator(rescale=1./255)
                                                                       In [4]:
x_train=train_datagen.flow_from_directory(r'/content/drive/MyDrive/Dataset
Plant Disease/fruit-dataset/fruit-
dataset/train',target size=(128,128),class mode='categorical',batch size=24
Found 5384 images belonging to 6 classes.
                                                                       In [5]:
x test=test datagen.flow from directory(r"/content/drive/MyDrive/Dataset
Plant Disease/fruit-dataset/fruit-dataset/test", target size=(128,128),
class mode='categorical',batch size=24)
Found 1686 images belonging to 6 classes.
Create the model
                                                                       In [6]:
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import
Dense, Convolution 2D, Max Pooling 2D, Flatten
Add Layers
                                                                       In [7]:
model=Sequential()
                                                                       In [9]:
model.add(Convolution2D(32,(3,3),input shape=(128,128,3),activation='relu')
                                                                      In [10]:
model.add(MaxPooling2D(pool size=(2,2)))
model.add(Flatten())
model.summary()
Model: "sequential"
Layer (type)
                            Output Shape
                                                     Param #
______
                             (None, 126, 126, 32)
 conv2d (Conv2D)
                                                      896
 max_pooling2d (MaxPooling2D (None, 63, 63, 32)
```

```
(None, 127008)
flatten (Flatten)
______
Total params: 896
Trainable params: 896
Non-trainable params: 0
                                                           In [11]:
32*(3*3*3+1)
model.add(Dense(300,activation='relu'))
model.add(Dense(150,activation='relu'))
                                                           In [12]:
model.add(Dense(6,activation='softmax'))
model.compile(loss='categorical crossentropy',optimizer='adam',metrics=['ac
curacy'l)
len(x train)
                                                          Out[12]:
225
                                                           In [13]:
1238/24
                                                          Out[13]:
51.583333333333336
fit the model
                                                           In [14]:
model.fit(x train, steps per epoch=len(x train), validation data=x test, valid
ation steps=len(x test),epochs=10)
Epoch 1/10
225/225 [============= ] - 2639s 12s/step - loss: 1.3354 -
accuracy: 0.7652 - val loss: 0.4907 - val accuracy: 0.8215
Epoch 2/10
accuracy: 0.9008 - val loss: 0.1736 - val accuracy: 0.9383
Epoch 3/10
225/225 [============ ] - 179s 794ms/step - loss: 0.2056 -
accuracy: 0.9296 - val loss: 0.1954 - val accuracy: 0.9312
Epoch 4/10
accuracy: 0.9383 - val loss: 0.2187 - val accuracy: 0.9253
Epoch 5/10
225/225 [=========== ] - 179s 796ms/step - loss: 0.1539 -
accuracy: 0.9461 - val loss: 0.1366 - val accuracy: 0.9543
Epoch 6/10
225/225 [=========== ] - 172s 765ms/step - loss: 0.1428 -
accuracy: 0.9491 - val loss: 0.1668 - val accuracy: 0.9442
Epoch 7/10
225/225 [=========== ] - 175s 774ms/step - loss: 0.1333 -
accuracy: 0.9538 - val loss: 0.1976 - val accuracy: 0.9253
Epoch 8/10
225/225 [============ ] - 174s 774ms/step - loss: 0.1172 -
accuracy: 0.9590 - val_loss: 0.0944 - val_accuracy: 0.9674
Epoch 9/10
225/225 [============ ] - 172s 763ms/step - loss: 0.1143 -
accuracy: 0.9569 - val loss: 0.1306 - val accuracy: 0.9561
```

```
Epoch 10/10
225/225 [============ ] - 179s 795ms/step - loss: 0.0913 -
accuracy: 0.9673 - val loss: 0.1848 - val accuracy: 0.9460
                                                                       Out[14]:
save the model
                                                                        In [16]:
model.save('fruitdata.h5')
Testing the model
                                                                        In [17]:
model=load model('fruitdata.h5')
                                                                        In [18]:
import numpy as np
from tensorflow.keras.models import load model
from tensorflow.keras.preprocessing import image
                                                                        In [19]:
img=image.load img(r"/content/drive/MyDrive/Dataset Plant Disease/fruit-
dataset/fruit-dataset/test/Apple___healthy/01efa999-757d-487e-8250-
27c7854c0ca8___RS_HL 7515.JPG",target_size=(128,128))
                                                                        In [21]:
img
                                                                        Out[21]:
                                                                        In [22]:
img=image.load img(r"/content/drive/MyDrive/Dataset Plant Disease/fruit-
dataset/fruit-dataset/test/Apple___healthy/01efa999-757d-487e-8250-
27c7854c0ca8___RS_HL 7515.JPG",target_size=(128,128))
                                                                        Out[22]:
                                                                        In [23]:
x=image.img_to_array(img)
                                                                        Out[23]:
array([[[111., 122., 152.],
        [110., 121., 151.],
        [118., 129., 159.],
        [184., 198., 225.],
```

```
[187., 201., 228.],
         [180., 194., 221.]],
       [[124., 135., 165.],
        [117., 128., 158.],
        [121., 132., 162.],
         . . . ,
        [178., 192., 219.],
         [191., 205., 232.],
         [179., 193., 220.]],
       [[123., 134., 164.],
        [117., 128., 158.],
        [114., 125., 155.],
        [180., 194., 221.],
        [189., 203., 230.],
        [179., 193., 220.]],
       . . . ,
       [[114., 128., 157.],
        [118., 132., 161.],
        [125., 139., 168.],
         . . . ,
        [177., 191., 217.],
        [187., 199., 223.],
        [180., 192., 216.]],
       [[120., 134., 163.],
        [125., 139., 168.],
        [122., 136., 165.],
         [189., 197., 216.],
         [188., 197., 214.],
         [186., 195., 210.]],
       [[127., 141., 170.],
        [118., 132., 161.],
        [118., 132., 161.],
        . . . ,
        [160., 167., 183.],
         [172., 180., 191.],
         [190., 199., 208.]]], dtype=float32)
                                                                           In [24]:
x=np.expand dims(x,axis=0)
                                                                          Out[24]:
array([[[[111., 122., 152.],
          [110., 121., 151.],
         [118., 129., 159.],
          [184., 198., 225.],
         [187., 201., 228.],
          [180., 194., 221.]],
```

```
[117., 128., 158.],
          [121., 132., 162.],
          [178., 192., 219.],
          [191., 205., 232.],
          [179., 193., 220.]],
         [[123., 134., 164.],
         [117., 128., 158.],
          [114., 125., 155.],
          . . . ,
          [180., 194., 221.],
          [189., 203., 230.],
          [179., 193., 220.]],
         . . . ,
         [[114., 128., 157.],
         [118., 132., 161.],
         [125., 139., 168.],
          [177., 191., 217.],
          [187., 199., 223.],
          [180., 192., 216.]],
         [[120., 134., 163.],
         [125., 139., 168.],
          [122., 136., 165.],
          . . . ,
         [189., 197., 216.],
          [188., 197., 214.],
          [186., 195., 210.]],
         [[127., 141., 170.],
         [118., 132., 161.],
         [118., 132., 161.],
          [160., 167., 183.],
          [172., 180., 191.],
          [190., 199., 208.]]]], dtype=float32)
                                                                            In [25]:
Х
                                                                           Out[25]:
array([[[[111., 122., 152.],
          [110., 121., 151.],
          [118., 129., 159.],
          . . . ,
          [184., 198., 225.],
          [187., 201., 228.],
          [180., 194., 221.]],
         [[124., 135., 165.],
         [117., 128., 158.],
          [121., 132., 162.],
          . . . ,
```

[[124., 135., 165.],

```
[178., 192., 219.],
         [191., 205., 232.],
         [179., 193., 220.]],
        [[123., 134., 164.],
         [117., 128., 158.],
         [114., 125., 155.],
         . . . ,
         [180., 194., 221.],
         [189., 203., 230.],
         [179., 193., 220.]],
        . . . ,
        [[114., 128., 157.],
         [118., 132., 161.],
         [125., 139., 168.],
         . . . ,
         [177., 191., 217.],
         [187., 199., 223.],
         [180., 192., 216.]],
        [[120., 134., 163.],
         [125., 139., 168.],
         [122., 136., 165.],
         [189., 197., 216.],
         [188., 197., 214.],
         [186., 195., 210.]],
        [[127., 141., 170.],
         [118., 132., 161.],
         [118., 132., 161.],
         [160., 167., 183.],
         [172., 180., 191.],
         [190., 199., 208.]]]], dtype=float32)
                                                                       In [26]:
y=np.argmax(model.predict(x),axis=1)
1/1 [=======] - 0s 201ms/step
                                                                       In [27]:
index=['Apple Black rot','Apple healthy','Corn (maize) Northern Leaf
Blight', 'Corn (maize) healthy', 'Peach Bacterial spot', 'Peach healthy'
index[y[0]]
                                                                      Out[27]:
'Apple healthy'
                                                                       In [28]:
img=image.load img(r"/content/drive/MyDrive/Dataset Plant Disease/fruit-
dataset/fruit-dataset/test/Apple healthy/01efa999-757d-487e-8250-
27c7854c0ca8 RS HL 7515.JPG", target size=(128,128))
x=image.img to array(img)
x=np.expand dims(x,axis=0)
y=np.argmax(model.predict(x),axis=1)
```