## import the libraries

```
In [18]:
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 [1]:
from tensorflow.keras.preprocessing.image import ImageDataGenerator
                                                                        In [2]:
train datagen=ImageDataGenerator(rescale=1./255,zoom range=0.2,horizontal f
lip=True, vertical flip=False)
                                                                        In [3]:
test datagen=ImageDataGenerator(rescale=1./255)
                                                                        In [4]:
x train=train datagen.flow from directory(r"/content/drive/MyDrive/Dataset
Plant Disease/Veg-dataset/Veg-dataset/train set", target size=(128,128),
class mode='categorical',batch size=24)
Found 11386 images belonging to 9 classes.
                                                                        In [5]:
x test=test datagen.flow from directory(r'/content/drive/MyDrive/Dataset
Plant Disease/Veg-dataset/Veg-
dataset/test set', target size=(128,128), class mode='categorical', batch size
=24)
Found 3416 images belonging to 9 classes.
add layers
                                                                        In [6]:
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import
Dense, Convolution 2D, Max Pooling 2D, Flatten
                                                                        In [7]:
model=Sequential()
                                                                        In [8]:
model.add(Convolution2D(32,(3,3),input shape=(128,128,3),activation='relu')
                                                                        In [9]:
model.add(MaxPooling2D(pool size=(2,2)))
                                                                       In [10]:
model.add(Flatten())
                                                                       In [11]:
model.summary()
Model: "sequential"
Layer (type)
                             Output Shape
                                                        Param #
______
```

```
conv2d (Conv2D)
                        (None, 126, 126, 32)
                                             896
max pooling2d (MaxPooling2D (None, 63, 63, 32)
                        (None, 127008)
flatten (Flatten)
______
Total params: 896
Trainable params: 896
Non-trainable params: 0
                                                           In [12]:
model.add(Dense(300,activation='relu'))
model.add(Dense(150,activation='relu'))
                                                           In [13]:
model.add(Dense(9,activation='softmax'))
                                                           In [14]:
model.compile(loss='categorical crossentropy',optimizer='adam',metrics=['ac
curacy'])
                                                           In [15]:
len(x train)
                                                          Out[15]:
475
                                                           In [16]:
1238/24
                                                          Out[16]:
51.583333333333336
fit the model
                                                           In [17]:
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
ccuracy: 0.6092 - val loss: 0.8031 - val accuracy: 0.7178
475/475 [============] - 375s 789ms/step - loss: 0.5598 -
accuracy: 0.8004 - val loss: 0.5123 - val accuracy: 0.8150
Epoch 3/10
accuracy: 0.8402 - val loss: 0.3696 - val accuracy: 0.8718
Epoch 4/10
475/475 [============= ] - 364s 765ms/step - loss: 0.3832 -
accuracy: 0.8669 - val loss: 0.2470 - val accuracy: 0.9183
Epoch 5/10
475/475 [============= ] - 364s 765ms/step - loss: 0.3210 -
accuracy: 0.8873 - val loss: 0.3565 - val accuracy: 0.8741
Epoch 6/10
475/475 [============= ] - 360s 758ms/step - loss: 0.2831 -
accuracy: 0.9040 - val loss: 0.4657 - val accuracy: 0.8352
Epoch 7/10
475/475 [============= ] - 371s 781ms/step - loss: 0.2453 -
accuracy: 0.9135 - val loss: 0.1700 - val accuracy: 0.9458
```

```
Epoch 8/10
475/475 [============= ] - 361s 760ms/step - loss: 0.2693 -
accuracy: 0.9062 - val loss: 0.3006 - val accuracy: 0.8967
accuracy: 0.9257 - val loss: 0.1678 - val accuracy: 0.9429
475/475 [============ ] - 373s 786ms/step - loss: 0.2082 -
accuracy: 0.9288 - val loss: 0.1483 - val accuracy: 0.9494
                                                              Out[17]:
save the model
                                                               In [19]:
model.save('vegetabledata.h5')
test the model
                                                               In [23]:
import numpy as np
from tensorflow.keras.models import load model
from tensorflow.keras.preprocessing import image
                                                               In [25]:
model=load model('vegetabledata.h5')
                                                               In [26]:
img=image.load_img(r"/content/drive/MyDrive/Dataset Plant Disease/Veg-
dataset/Veg-dataset/test_set/Potato___Early_blight/b7883606-5157-4dc1-b965-
fc10f8fe1796 RS Early.B 7598.JPG")
img
```





In [27]:

x=image.img\_to\_array(img)
img=image.load\_img(r"/content/drive/MyDrive/Dataset Plant Disease/Vegdataset/Veg-dataset/test\_set/Potato\_\_\_Early\_blight/b7883606-5157-4dc1-b965fc10f8fe1796\_\_\_RS\_Early.B 7598.JPG",target\_size=(128,128))
img

Out[27]:



```
x=image.img_to_array(img)
array([[[156., 163., 191.],
        [158., 165., 193.],
        [155., 162., 190.],
        [109., 113., 140.],
        [109., 113., 140.],
        [113., 117., 144.]],
       [[166., 173., 201.],
        [166., 173., 201.],
        [159., 166., 194.],
        . . . ,
        [110., 114., 141.],
        [104., 108., 135.],
        [109., 113., 140.]],
       [[168., 175., 203.],
        [160., 167., 195.],
        [152., 159., 187.],
        [110., 114., 141.],
        [101., 105., 132.],
        [110., 114., 141.]],
       . . . ,
       [[160., 161., 181.],
        [162., 163., 183.],
        [155., 156., 176.],
        [103., 101., 122.],
        [ 99., 97., 118.],
        [105., 103., 124.]],
       [[155., 156., 176.],
        [150., 151., 171.],
        [152., 153., 173.],
        [109., 107., 128.],
        [102., 100., 121.],
        [107., 105., 126.]],
       [[157., 158., 178.],
```

[156., 157., 177.],

In [28]:

Out[28]:

```
[149., 150., 170.],
         . . . ,
         [ 92., 90., 111.],
        [119., 117., 138.],
         [ 96., 94., 115.]]], dtype=float32)
                                                                           In [29]:
x=np.expand dims(x,axis=0)
                                                                           In [30]:
Х
                                                                          Out[30]:
array([[[[156., 163., 191.],
          [158., 165., 193.],
          [155., 162., 190.],
          . . . ,
          [109., 113., 140.],
          [109., 113., 140.],
         [113., 117., 144.]],
         [[166., 173., 201.],
         [166., 173., 201.],
         [159., 166., 194.],
         [110., 114., 141.],
         [104., 108., 135.],
         [109., 113., 140.]],
         [[168., 175., 203.],
         [160., 167., 195.],
         [152., 159., 187.],
          [110., 114., 141.],
          [101., 105., 132.],
         [110., 114., 141.]],
         . . . ,
         [[160., 161., 181.],
         [162., 163., 183.],
         [155., 156., 176.],
         [103., 101., 122.],
          [ 99., 97., 118.],
         [105., 103., 124.]],
         [[155., 156., 176.],
         [150., 151., 171.],
         [152., 153., 173.],
          [109., 107., 128.],
         [102., 100., 121.],
         [107., 105., 126.]],
         [[157., 158., 178.],
         [156., 157., 177.],
          [149., 150., 170.],
```

```
. . . ,
        [ 92., 90., 111.],
        [119., 117., 138.],
         [ 96., 94., 115.]]], dtype=float32)
                                                                    In [31]:
y=np.argmax(model.predict(x),axis=1)
1/1 [=======] - 0s 208ms/step
                                                                    In [32]:
x train.class indices
                                                                   Out[32]:
{'Pepper, bell Bacterial spot': 0,
 'Pepper, bell healthy': 1,
 'Potato___Early_blight': 2,
 'Potato___Late_blight': 3,
 'Potato___healthy': 4,
 'Tomato___Bacterial_spot': 5,
 'Tomato Late blight': 6,
 'Tomato Leaf Mold': 7,
 'Tomato Septoria leaf spot': 8}
                                                                    In [35]:
index=['Pepper, bell Bacterial spot','Pepper, bell healthy','Potato E
arly blight', 'Potato Late blight', 'Potato healthy', 'Tomato Bacterial
spot','Tomato___Late_blight','Tomato___Leaf_Mold','Tomato___Septoria_leaf_s
pot']
                                                                    In [36]:
index[y[0]]
                                                                   Out[36]:
'Potato Early_blight'
                                                                    In [37]:
img=image.load_img(r"/content/drive/MyDrive/Dataset Plant Disease/Veg-
dataset/Veg-dataset/test set/Potato Early blight/b7883606-5157-4dc1-b965-
fc10f8fe1796 RS Early.B 7598.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)
index=['Pepper,_bell___Bacterial_spot','Pepper,_bell___healthy','Potato___E
arly blight', 'Potato Late blight', 'Potato healthy', 'Tomato Bacterial
spot','Tomato___Leaf_Mold','Tomato___Septoria_leaf_spot']
index[y[0]]
1/1 [======== ] - 0s 60ms/step
                                                                   Out[37]:
'Potato Early blight'
```