```
import numpy as np
 In [1]:
         import tensorflow as tf
         from keras.preprocessing.image import ImageDataGenerator
 In [2]: train_datagen = ImageDataGenerator(
                 rescale=1./255,
                 shear_range=0.2,
                 zoom_range=0.2,
                 horizontal_flip=True)
         training_set = train_datagen.flow_from_directory(
                  'training set',
                 target size=(64, 64),
                 batch size=32,
                 class_mode='categorical')
         Found 1028 images belonging to 8 classes.
         test_datagen = ImageDataGenerator(rescale=1./255)
 In [3]:
         test set = test datagen.flow from directory(
                  'test_set',
                 target size=(64, 64),
                 batch size=32,
                 class_mode='categorical')
         Found 394 images belonging to 8 classes.
 In [4]: cnn = tf.keras.models.Sequential()
 In [5]:
         cnn.add(tf.keras.layers.Conv2D(filters=64 , kernel size=3 , activation='relu' , in
         cnn.add(tf.keras.layers.MaxPool2D(pool size=2,strides=2))
         cnn.add(tf.keras.layers.Conv2D(filters=64 , kernel_size=3 , activation='relu' ))
 In [6]:
          cnn.add(tf.keras.layers.MaxPool2D(pool_size=2 , strides=2))
 In [7]:
         cnn.add(tf.keras.layers.Dropout(0.5))
         cnn.add(tf.keras.layers.Flatten())
In [8]:
         cnn.add(tf.keras.layers.Dense(units=128, activation='relu'))
In [9]:
         cnn.add(tf.keras.layers.Dense(units=8 , activation='softmax'))
In [10]:
         cnn.compile(optimizer = 'rmsprop' , loss = 'categorical_crossentropy' , metrics =
In [11]:
         cnn.fit(x = training set , validation data = test set , epochs = 30)
In [12]:
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Epoch 1/30
33/33 [=============== ] - 168s 5s/step - loss: 1.8029 - accuracy:
0.3677 - val_loss: 1.2824 - val_accuracy: 0.4670
Epoch 2/30
0.5846 - val loss: 0.8779 - val accuracy: 0.6701
Epoch 3/30
33/33 [================] - 132s 4s/step - loss: 0.9449 - accuracy:
0.6887 - val loss: 0.6170 - val accuracy: 0.7893
Epoch 4/30
33/33 [================] - 144s 4s/step - loss: 0.7436 - accuracy:
0.7471 - val_loss: 0.3967 - val_accuracy: 0.8756
Epoch 5/30
0.7821 - val loss: 0.4915 - val accuracy: 0.8223
Epoch 6/30
0.8045 - val_loss: 0.2742 - val_accuracy: 0.9061
Epoch 7/30
33/33 [==================] - 121s 4s/step - loss: 0.4994 - accuracy:
0.8259 - val_loss: 0.2895 - val_accuracy: 0.9112
Epoch 8/30
33/33 [================= ] - 136s 4s/step - loss: 0.4838 - accuracy:
0.8453 - val loss: 0.2649 - val accuracy: 0.9112
Epoch 9/30
33/33 [================= ] - 135s 4s/step - loss: 0.4354 - accuracy:
0.8541 - val_loss: 0.3494 - val_accuracy: 0.8909
Epoch 10/30
0.8755 - val_loss: 0.2119 - val_accuracy: 0.9213
Epoch 11/30
33/33 [================== ] - 145s 4s/step - loss: 0.3468 - accuracy:
0.8842 - val_loss: 0.7019 - val_accuracy: 0.7614
Epoch 12/30
33/33 [=================] - 139s 4s/step - loss: 0.3202 - accuracy:
0.8862 - val_loss: 0.1615 - val_accuracy: 0.9543
Epoch 13/30
0.8833 - val loss: 0.1003 - val accuracy: 0.9645
Epoch 14/30
33/33 [=================] - 144s 4s/step - loss: 0.2960 - accuracy:
0.8901 - val loss: 0.1006 - val accuracy: 0.9772
Epoch 15/30
0.9202 - val loss: 0.1135 - val accuracy: 0.9670
Epoch 16/30
33/33 [================] - 141s 4s/step - loss: 0.2435 - accuracy:
0.9144 - val loss: 0.1164 - val accuracy: 0.9670
Epoch 17/30
33/33 [================] - 144s 4s/step - loss: 0.2341 - accuracy:
0.9212 - val_loss: 0.1013 - val_accuracy: 0.9695
Epoch 18/30
33/33 [================] - 129s 4s/step - loss: 0.1870 - accuracy:
0.9329 - val_loss: 0.1845 - val_accuracy: 0.9340
Epoch 19/30
33/33 [================= ] - 141s 4s/step - loss: 0.1831 - accuracy:
0.9377 - val_loss: 0.1308 - val_accuracy: 0.9391
Epoch 20/30
33/33 [================= ] - 136s 4s/step - loss: 0.1881 - accuracy:
0.9319 - val loss: 0.0385 - val accuracy: 0.9898
Epoch 21/30
33/33 [==============] - 130s 4s/step - loss: 0.2194 - accuracy:
0.9358 - val_loss: 0.0686 - val_accuracy: 0.9797
Epoch 22/30
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33/33 [============= ] - 138s 4s/step - loss: 0.1719 - accuracy:
       0.9368 - val_loss: 0.1827 - val_accuracy: 0.9213
       Epoch 23/30
       33/33 [================] - 152s 5s/step - loss: 0.2468 - accuracy:
       0.9202 - val loss: 0.0425 - val accuracy: 0.9924
       Epoch 24/30
       33/33 [================] - 179s 5s/step - loss: 0.1418 - accuracy:
       0.9553 - val loss: 0.0688 - val accuracy: 0.9822
       Epoch 25/30
       33/33 [================] - 173s 5s/step - loss: 0.1490 - accuracy:
       0.9572 - val_loss: 0.0657 - val_accuracy: 0.9695
       Epoch 26/30
       0.9484 - val loss: 0.0744 - val accuracy: 0.9772
       Epoch 27/30
       0.9475 - val_loss: 0.0597 - val_accuracy: 0.9797
       Epoch 28/30
       0.9446 - val_loss: 0.0417 - val_accuracy: 0.9873
       Epoch 29/30
       33/33 [================] - 166s 5s/step - loss: 0.1163 - accuracy:
       0.9621 - val_loss: 0.0614 - val_accuracy: 0.9797
       Epoch 30/30
       0.9621 - val_loss: 0.0267 - val_accuracy: 0.9949
       <keras.callbacks.History at 0x24839e4c220>
Out[12]:
In [13]:
       cnn.save("leafindt.h5")
       from tensorflow import keras
In [14]:
       model = keras.models.load_model('leafindt.h5')
In [20]:
       from keras preprocessing import image
       import numpy as np
       prediction_img= image.load_img('Predection/11.jpg', target_size=(64,64))
       prediction_img = image.img_to_array(prediction_img)
       prediction_img = np.expand_dims(prediction_img,axis=0)
       # predicting image
       result = model.predict(prediction_img)
       1/1 [======] - 2s 2s/step
In [21]: print(result)
       [[0. 0. 0. 0. 0. 0. 1. 0.]]
In [23]:
       if result[0][0]==1:
           print('PAN')
       elif result[0][1]==1:
          print('GUAVA')
       elif result[0][2]==1:
           print('JOBA')
       elif result[0][3]==1:
           print('LEMON')
       elif result[0][4]==1:
           print('MANGO')
       elif result[0][5]==1:
           print('MINT')
       elif result[0][6]==1:
           print('NEEM')
```

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elif result[0][7]==1:
  print('TULSI')
```

NEEM