

```

from tensorflow.keras.preprocessing.image import ImageDataGenerator

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

test_datagen=ImageDataGenerator(rescale=1./255)

x_train=train_datagen.flow_from_directory('fruit-dataset/f_train',
                                          target_size=(128,128), batch_size=24, class_mode='categorical')

x_test=test_datagen.flow_from_directory('fruit-dataset/f_test',
                                       target_size=(128,128), batch_size=24, class_mode='categorical')

```

Found 5384 images belonging to 6 classes.

Found 1686 images belonging to 6 classes.

x\_train.class\_indices

```

{'Apple___Black_rot': 0,
 'Apple___healthy': 1,
 'Corn_(maize)___Northern_Leaf_Blight': 2,
 'Corn_(maize)___healthy': 3,
 'Peach___Bacterial_spot': 4,
 'Peach___healthy': 5}

```

```

from tensorflow.keras.models import Sequential

```

```

from tensorflow.keras.layers import Dense, Convolution2D, MaxPooling2D, Flatten

```

```

model=Sequential()

```

```

model.add(Convolution2D(32,(3,3),input_shape=(128,128,3),activation='relu'))

```

```

model.add(MaxPooling2D(pool_size=(2,2)))

```

```

model.add(Flatten())

```

```

model.summary()

```

Model: "sequential"

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Layer (type)	Output Shape	Param #
=====		
conv2d (Conv2D)	(None, 126, 126, 32)	896
max_pooling2d (MaxPooling2D)	(None, 63, 63, 32)	0

)

flatten (Flatten) (None, 127008) 0

=====

Total params: 896

Trainable params: 896

Non-trainable params: 0

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

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

```
model.add(Dense(6,activation='softmax'))
```

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

```
model.fit(x_train,steps_per_epoch=len(x_train),validation_data=x_test,validation_steps=len(x_test),
epochs=10)
```

Epoch 1/10

225/225 [=====] - 158s 701ms/step - loss: 1.1387 - accuracy: 0.7624 -  
val\_loss: 0.3235 - val\_accuracy: 0.8843

Epoch 2/10

225/225 [=====] - 111s 491ms/step - loss: 0.2539 - accuracy: 0.9116 -  
val\_loss: 0.2671 - val\_accuracy: 0.9039

Epoch 3/10

225/225 [=====] - 118s 525ms/step - loss: 0.1975 - accuracy: 0.9329 -  
val\_loss: 0.1724 - val\_accuracy: 0.9478

Epoch 4/10

225/225 [=====] - 127s 565ms/step - loss: 0.1653 - accuracy: 0.9421 -  
val\_loss: 0.2982 - val\_accuracy: 0.9004

Epoch 5/10

225/225 [=====] - 127s 565ms/step - loss: 0.1533 - accuracy: 0.9461 -  
val\_loss: 0.1494 - val\_accuracy: 0.9531

Epoch 6/10

225/225 [=====] - 113s 500ms/step - loss: 0.1154 - accuracy: 0.9573 -  
val\_loss: 0.1461 - val\_accuracy: 0.9520

Epoch 7/10

225/225 [=====] - 114s 508ms/step - loss: 0.1154 - accuracy: 0.9625 -  
val\_loss: 0.2850 - val\_accuracy: 0.9211

Epoch 8/10

225/225 [=====] - 112s 496ms/step - loss: 0.1047 - accuracy: 0.9627 -  
val\_loss: 0.1464 - val\_accuracy: 0.9591

Epoch 9/10

225/225 [=====] - 107s 475ms/step - loss: 0.1242 - accuracy: 0.9567 -  
val\_loss: 0.1058 - val\_accuracy: 0.9656

Epoch 10/10

225/225 [=====] - 106s 470ms/step - loss: 0.1006 - accuracy: 0.9651 -  
val\_loss: 0.2158 - val\_accuracy: 0.9312

score,acc=model.evaluate(x\_test,batch\_size=128,verbose=2)

acc

71/71 - 5s - loss: 0.2158 - accuracy: 0.9312 - 5s/epoch - 67ms/step

0.9311981201171875

model.save('fruit.h5')