

Fruit disease predictions

Volume in drive C is Windows-SSD

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Directory of

C:\Users\maris_q3mm6nk\Desktop\FILES\data_for_ibm\Fertilizers_Recommendation_System_For_Disease_Prediction\Dataset Plant Disease

```
22-10-22 10:33 AM      .
28-09-22 08:07 PM      ..
22-10-22 10:03 AM      .ipynb_checkpoints
28-09-22 08:07 PM      fruit-dataset
22-10-22 10:33 AM      5,899 Untitled.ipynb
28-09-22 08:08 PM      Veg-dataset
          1 File(s)      5,899 bytes
          5 Dir(s) 160,126,849,024 bytes free
```

pwd

'C:\\Users\\maris_q3mm6nk\\Desktop\\FILES\\data_for_ibm\\Fertilizers_Recommendation_System_For_Disease_Prediction\\Dataset Plant Disease'

I

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)

ls

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```
22-10-22 10:33 AM      .
28-09-22 08:07 PM      ..
22-10-22 10:03 AM      .ipynb_checkpoints
28-09-22 08:07 PM      fruit-dataset
22-10-22 10:33 AM      5,899 Untitled.ipynb
28-09-22 08:08 PM      Veg-dataset
```

1 File(s) 5,899 bytes
5 Dir(s) 160,126,529,536 bytes free

```
x_train=train_datagen.flow_from_directory(r"C:\Users\Praveen\Desktop\FILES\data_for_ibm\Fertilizers_Recommendation_System_For_Disease_Prediction\Dataset Plant Disease\fruit-dataset\fruit-dataset\train",target_size=(128,128),
```

```
x_test=test_datagen.flow_from_directory(r"C:\Users\Praveen\Desktop\FILES\data_for_ibm\Fertilizers_Recommendation_System_For_Disease_Prediction\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.

```
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"

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		

```
32*(3*3*3+1)
```

```
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'])
len(x_train)
```

225

1238/24

51.583333333333336

```
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 [=====] - 125s 554ms/step - loss: 0.0932 - accuracy: 0.9690 - val_loss: 0.1116 - val_accuracy: 0.9632

Epoch 2/10

225/225 [=====] - 125s 555ms/step - loss: 0.0797 - accuracy: 0.9762 - val_loss: 0.2585 - val_accuracy: 0.9306

Epoch 3/10

225/225 [=====] - 126s 561ms/step - loss: 0.0734 - accuracy: 0.9734 - val_loss: 0.1670 - val_accuracy: 0.9537

Epoch 4/10

225/225 [=====] - 126s 560ms/step - loss: 0.0613 - accuracy: 0.9785 - val_loss: 0.0807 - val_accuracy: 0.9745

Epoch 5/10

225/225 [=====] - 120s 533ms/step - loss: 0.0713 - accuracy: 0.9733 - val_loss: 0.0947 - val_accuracy: 0.9674

Epoch 6/10

225/225 [=====] - 117s 521ms/step - loss: 0.0655 - accuracy: 0.9759 - val_loss: 0.0663 - val_accuracy: 0.9757

Epoch 7/10

225/225 [=====] - 120s 535ms/step - loss: 0.0518 - accuracy: 0.9807 - val_loss: 0.1740 - val_accuracy: 0.9531

Epoch 8/10

225/225 [=====] - 108s 478ms/step - loss: 0.0579 - accuracy: 0.9786 - val_loss: 0.1072 - val_accuracy: 0.9727

Epoch 9/10

225/225 [=====] - 105s 467ms/step - loss: 0.0530 - accuracy: 0.9824 - val_loss: 0.0768 - val_accuracy: 0.9763

Epoch 10/10

225/225 [=====] - 114s 507ms/step - loss: 0.0692 - accuracy: 0.9779 - val_loss: 0.1067 - val_accuracy: 0.9614

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

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

```
model=load_model('fruitdata.h5')
```

```
img=image.load_img(r"C:\Users\Praveen\Desktop\FILES\data_for_ibm\Fertilizers_Recommendation_System_For_Disease_Prediction\Dataset Plant Disease\fruit-dataset\fruit-dataset\test\Apple___healthy\00fca0da-2db3-481b-b98a-9b67bb7b105c___RS_HL_7708.jpg")
```

```
img
```

```
img=image.load_img(r"C:\Users\Praveen\Desktop\FILES\data_for_ibm\Fertilizers_Recommendation_System_For_Disease_Prediction\Dataset Plant Disease\fruit-dataset\fruit-dataset\test\Apple___healthy\00fca0da-2db3-481b-b98a-9b67bb7b105c___RS_HL_7708.jpg",target_size=(128,128))
img
```

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

```
array([[[165., 153., 189.],
        [165., 153., 189.],
        [165., 153., 189.],
        ...,
        [176., 170., 206.],
        [176., 170., 206.],
        [176., 170., 206.]],

       [[164., 152., 188.],
        [164., 152., 188.],
        [164., 152., 188.],
        ...,
        [173., 167., 203.],
        [172., 166., 202.],
        [172., 166., 202.]])
```

```

[[163., 151., 187.],
 [163., 151., 187.],
 [163., 151., 187.],
 ...,
 [172., 166., 202.],
 [170., 164., 200.],
 [169., 163., 199.]],

...,

[[135., 119., 156.],
 [139., 123., 160.],
 [134., 118., 155.],
 ...,
 [143., 133., 168.],
 [138., 128., 163.],
 [141., 131., 166.]],

[[136., 120., 157.],
 [134., 118., 155.],
 [134., 118., 155.],
 ...,
 [141., 131., 166.],
 [141., 131., 166.],
 [146., 136., 171.]],

[[135., 119., 156.],
 [140., 124., 161.],
 [143., 127., 164.],
 ...,
 [145., 135., 170.],
 [151., 141., 176.],
 [140., 130., 165.]]], dtype=float32)

```

```

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

```

```

array([[[[165., 153., 189.],
 [165., 153., 189.],
 [165., 153., 189.],
 ...,
 [176., 170., 206.],
 [176., 170., 206.],
 [176., 170., 206.]],

[[164., 152., 188.],
 [164., 152., 188.],

```

```

[164., 152., 188.],
...,
[173., 167., 203.],
[172., 166., 202.],
[172., 166., 202.]],

[[163., 151., 187.],
[163., 151., 187.],
[163., 151., 187.],
...,
[172., 166., 202.],
[170., 164., 200.],
[169., 163., 199.]],

...,

[[135., 119., 156.],
[139., 123., 160.],
[134., 118., 155.],
...,
[143., 133., 168.],
[138., 128., 163.],
[141., 131., 166.]],

[[136., 120., 157.],
[134., 118., 155.],
[134., 118., 155.],
...,
[141., 131., 166.],
[141., 131., 166.],
[146., 136., 171.]],

[[135., 119., 156.],
[140., 124., 161.],
[143., 127., 164.],
...,
[145., 135., 170.],
[151., 141., 176.],
[140., 130., 165.]]], dtype=float32)

```

x

```

array([[[[165., 153., 189.],
[165., 153., 189.],
[165., 153., 189.],
...,

```

```

[176., 170., 206.],
[176., 170., 206.],
[176., 170., 206.]],

[[164., 152., 188.],
 [164., 152., 188.],
 [164., 152., 188.],
 ...,
 [173., 167., 203.],
 [172., 166., 202.],
 [172., 166., 202.]],

[[163., 151., 187.],
 [163., 151., 187.],
 [163., 151., 187.],
 ...,
 [172., 166., 202.],
 [170., 164., 200.],
 [169., 163., 199.]],

...,

[[135., 119., 156.],
 [139., 123., 160.],
 [134., 118., 155.],
 ...,
 [143., 133., 168.],
 [138., 128., 163.],
 [141., 131., 166.]],

[[136., 120., 157.],
 [134., 118., 155.],
 [134., 118., 155.],
 ...,
 [141., 131., 166.],
 [141., 131., 166.],
 [146., 136., 171.]],

[[135., 119., 156.],
 [140., 124., 161.],
 [143., 127., 164.],
 ...,
 [145., 135., 170.],
 [151., 141., 176.],
 [140., 130., 165.]]], dtype=float32)

```

```
y=np.argmax(model.predict(x),axis=1)
```

```
1/1 [=====] - 0s 71ms/step
```

```
x_train.class_indice
```

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

```
index=['Apple___Black_rot','Apple___healthy','Corn_(maize)___Northern_Leaf_Blight','Corn_(  
maize)___healthy','Peach___Bacterial_spot','Peach___healthy']
```

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

```
'Apple___healthy'
```

```
img=image.load_img(r"C:\Users\Praveen\Desktop\FILES\data_for_ibm\Fertilizers_Recommend  
ation_System_For_Disease_Prediction\Dataset Plant Disease\fruit-dataset\fruit-  
dataset\test\Apple___healthy\00fca0da-2db3-481b-b98a-9b67bb7b105c___RS_HL  
7708.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=['Apple___Black_rot','Apple___healthy','Corn_(maize)___Northern_Leaf_Blight','Corn_(  
maize)___healthy','Peach___Bacterial_spot','Peach___healthy']
```

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

```
1/1 [=====] - 0s 33ms/step
```

```
'Apple___healthy'
```