```
In [1]: import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
         import seaborn as sns
         import tensorflow as tf
In [2]:
         import os
         import cv2 as cv
In [3]: from tensorflow import keras
In [4]: from keras import layers, Sequential ,Model
 In [5]: from keras.layers import Flatten,Dropout,Dense,Input,Lambda,MaxPooling3D,Conv2D
 In [6]: path=r"C:\Users\HP\Downloads\banana\BananaLSD\OriginalSet"
 In [7]: | dis=os.listdir(path)
In [8]: | dis
Out[8]: ['cordana', 'healthy', 'pestalotiopsis', 'sigatoka']
In [9]: Data=[]
In [10]: for i in dis:
             A=os.path.join(path,i)
             for j in os.listdir(A):
                 B=os.path.join(A,j)
                 img=cv.imread(B)
                 C=cv.resize(img,(130,130))
                 T=dis.index(i)
                 Data.append([C,T])
         import random
In [11]:
         random.shuffle(Data)
In [12]: T=dis.index('cordana')
         print(T)
```

```
In [13]: F=[]
         T=[]
         for i,j in Data:
             F.append(i)
             T.append(j)
In [14]: F=np.array(F)
In [15]: T=np.array(T)
In [16]: F=F/25
In [17]: T=pd.get_dummies(T,dtype="int")
         VGG16
In [18]: from keras.applications.vgg16 import VGG16
In [19]: from keras.applications.vgg16 import preprocess_input
In [20]: vgg=VGG16(input_shape=(130,130,3),weights='imagenet',include_top=False)
In [21]: for i in vgg.layers:
             i.trainabler=False
In [22]: X=Flatten()(vgg.output)
In [23]: y=Dense(20,activation='relu')(X)
In [24]: z=Dense(4,activation='softmax')(y)
In [25]: model=Model(vgg.input,z)
In [26]: from keras.losses import CategoricalCrossentropy
In [27]: | model.compile(optimizer="adam",
                      loss="CategoricalCrossentropy",
                      metrics=["accuracy"])
```

In [28]: model.fit(F,T,epochs=20,validation_split=0.15)

```
Epoch 1/20
25/25 ——
                 54s 2s/step - accuracy: 0.4005 - loss: 2.2554 - val accur
acy: 0.5319 - val loss: 1.2082
Epoch 2/20
                  52s 2s/step - accuracy: 0.4982 - loss: 1.1819 - val_accur
25/25 ----
acy: 0.5390 - val loss: 0.9672
Epoch 3/20
25/25 -
                        - 54s 2s/step - accuracy: 0.5334 - loss: 0.9057 - val_accur
acy: 0.5532 - val_loss: 0.8324
Epoch 4/20
                        - 52s 2s/step - accuracy: 0.5847 - loss: 0.8526 - val accur
25/25 -
acy: 0.4397 - val loss: 1.2445
Epoch 5/20
                53s 2s/step - accuracy: 0.5846 - loss: 1.0099 - val_accur
25/25 ——
acy: 0.6738 - val loss: 0.8477
Epoch 6/20
               _______ 52s 2s/step - accuracy: 0.6699 - loss: 0.7863 - val accur
25/25 ———
acy: 0.7021 - val_loss: 0.7412
Epoch 7/20
                      52s 2s/step - accuracy: 0.7216 - loss: 0.7038 - val_accur
25/25 -
acy: 0.6879 - val_loss: 0.7326
Epoch 8/20
25/25 -
                       - 52s 2s/step - accuracy: 0.7003 - loss: 0.7131 - val accur
acy: 0.7305 - val loss: 0.6671
Epoch 9/20
25/25 ——
               ________ 53s 2s/step - accuracy: 0.7104 - loss: 0.6928 - val_accur
acy: 0.6667 - val loss: 0.6346
Epoch 10/20
                       - 53s 2s/step - accuracy: 0.7082 - loss: 0.6534 - val accur
25/25 -
acy: 0.7376 - val loss: 0.6100
Epoch 11/20
25/25 -
                     54s 2s/step - accuracy: 0.7724 - loss: 0.6027 - val accur
acy: 0.6950 - val_loss: 0.7253
Epoch 12/20
                       -- 55s 2s/step - accuracy: 0.7315 - loss: 0.7275 - val accur
25/25 -
acy: 0.7163 - val loss: 0.6042
Epoch 13/20
25/25 ———
               _______ 55s 2s/step - accuracy: 0.7380 - loss: 0.6019 - val_accur
acy: 0.7801 - val loss: 0.6060
Epoch 14/20
25/25 -
                      56s 2s/step - accuracy: 0.7935 - loss: 0.5361 - val accur
acy: 0.7376 - val_loss: 0.6290
Epoch 15/20
25/25 -
                      -- 56s 2s/step - accuracy: 0.8334 - loss: 0.4767 - val accur
acy: 0.7730 - val loss: 0.6996
Epoch 16/20
25/25 -
                        - 55s 2s/step - accuracy: 0.8308 - loss: 0.4663 - val_accur
acy: 0.7376 - val loss: 0.6625
Epoch 17/20
25/25 ———
                 _______ 55s 2s/step - accuracy: 0.7751 - loss: 0.5768 - val_accur
acy: 0.7305 - val_loss: 0.9035
Epoch 18/20
                      55s 2s/step - accuracy: 0.8091 - loss: 0.5119 - val accur
acy: 0.7730 - val loss: 0.7505
Epoch 19/20
25/25 -
                        - 55s 2s/step - accuracy: 0.8111 - loss: 0.4996 - val_accur
acy: 0.7872 - val loss: 0.6307
Epoch 20/20
```

```
25/25 ________ 55s 2s/step - accuracy: 0.7925 - loss: 0.5100 - val_accur acy: 0.7872 - val_loss: 0.9144

Out[28]: <keras.src.callbacks.history.History at 0x13e5d7a85d0>

In [90]: img=r"C:\Users\HP\Downloads\banana\BananaLSD\OriginalSet\sigatoka\5.jpeg"

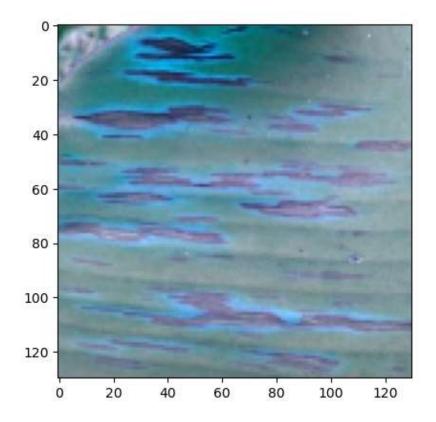
In [91]: img1=cv.imread(img)

In [92]: img2=cv.resize(img1,(130,130))

In [93]: img31=img2.reshape(1,130,130,3)

In [94]: plt.imshow(img2)
```

Out[94]: <matplotlib.image.AxesImage at 0x13eec51b390>



Out[96]: 3

VGG19

In [46]: model.fit(F,T,epochs=20,validation_split=0.15)

```
Epoch 1/20
25/25 ———
               73s 3s/step - accuracy: 0.3825 - loss: 5.1380 - val_accur
acy: 0.5532 - val loss: 0.9374
Epoch 2/20
                  71s 3s/step - accuracy: 0.5313 - loss: 0.9899 - val_accur
25/25 ----
acy: 0.5674 - val loss: 0.8830
Epoch 3/20
25/25 -
                       - 70s 3s/step - accuracy: 0.5550 - loss: 0.9217 - val_accur
acy: 0.5887 - val_loss: 0.8596
Epoch 4/20
                       - 72s 3s/step - accuracy: 0.6041 - loss: 0.9165 - val accur
25/25 -
acy: 0.5461 - val loss: 0.8855
Epoch 5/20
               70s 3s/step - accuracy: 0.5819 - loss: 0.9033 - val_accur
25/25 ——
acy: 0.6312 - val loss: 0.8009
Epoch 6/20
               70s 3s/step - accuracy: 0.6159 - loss: 0.9088 - val accur
25/25 ———
acy: 0.6170 - val_loss: 0.7900
Epoch 7/20
                      — 69s 3s/step - accuracy: 0.6276 - loss: 0.8117 - val_accur
25/25 -
acy: 0.7234 - val_loss: 0.7246
Epoch 8/20
25/25 -
                       - 69s 3s/step - accuracy: 0.6178 - loss: 0.8143 - val accur
acy: 0.5957 - val loss: 0.8203
Epoch 9/20
               70s 3s/step - accuracy: 0.5882 - loss: 0.8613 - val_accur
25/25 ——
acy: 0.6525 - val loss: 0.7306
Epoch 10/20
                       - 72s 3s/step - accuracy: 0.6352 - loss: 0.8277 - val accur
25/25 -
acy: 0.6312 - val loss: 0.7413
Epoch 11/20
25/25 -
                    71s 3s/step - accuracy: 0.6042 - loss: 0.7981 - val accur
acy: 0.6809 - val_loss: 0.6839
Epoch 12/20
                       - 71s 3s/step - accuracy: 0.6193 - loss: 0.8511 - val accur
25/25 -
acy: 0.6596 - val loss: 0.7162
Epoch 13/20
25/25 ———
               70s 3s/step - accuracy: 0.6469 - loss: 0.7930 - val accur
acy: 0.7234 - val loss: 0.5743
Epoch 14/20
25/25 -
                     70s 3s/step - accuracy: 0.6846 - loss: 0.6941 - val accur
acy: 0.7376 - val_loss: 0.5724
Epoch 15/20
25/25 -
                      -- 73s 3s/step - accuracy: 0.7043 - loss: 0.6282 - val accur
acy: 0.7589 - val loss: 0.4965
Epoch 16/20
25/25 -
                       - 71s 3s/step - accuracy: 0.7193 - loss: 0.5547 - val_accur
acy: 0.7163 - val loss: 0.6175
Epoch 17/20
25/25 ———
               70s 3s/step - accuracy: 0.7084 - loss: 0.5551 - val accur
acy: 0.6667 - val loss: 0.7286
Epoch 18/20
                     69s 3s/step - accuracy: 0.6725 - loss: 0.6653 - val accur
25/25 ———
acy: 0.7660 - val loss: 0.4877
Epoch 19/20
25/25 -
                        - 75s 3s/step - accuracy: 0.7440 - loss: 0.6033 - val_accur
acy: 0.8014 - val loss: 0.4482
Epoch 20/20
```

```
25/25 91s 4s/step - accuracy: 0.7449 - loss: 0.5392 - val_accuracy: 0.7376 - val_loss: 0.5823

Out[46]: <a href="mailto:keras.src.callbacks.history.History">keras.src.callbacks.history.History</a> at 0x13e00859f50>

In [97]: img=r"C:\Users\HP\Downloads\banana\BananaLSD\OriginalSet\pestalotiopsis\84.jpeg"

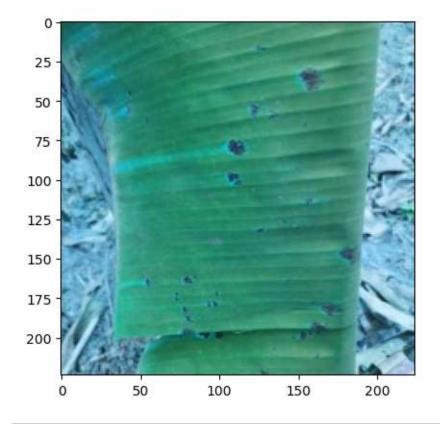
In [98]: img1=cv.imread(img)

In [99]: img2=cv.resize(img1,(130,130))

In [100]: img3=img2.reshape(1,130,130,3)

In [101]: plt.imshow(img1)
```

Out[101]: <matplotlib.image.AxesImage at 0x13eef8b00d0>



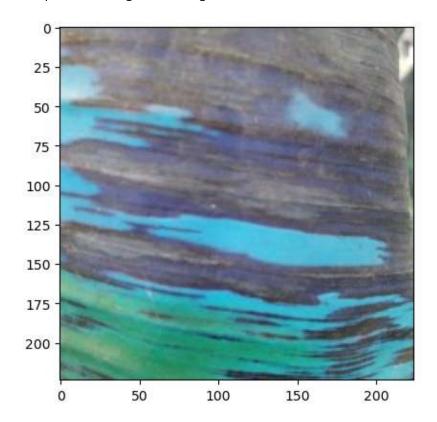
Resnet

```
In [64]: model.fit(F,T,epochs=20,validation_split=0.15)
          Epoch 1/10
          25/25 -
                                  -- 76s 2s/step - accuracy: 0.5902 - loss: 2.6507 - val accur
          acy: 0.5319 - val_loss: 3690005.0000
          Epoch 2/10
          25/25 -
                                    - 44s 2s/step - accuracy: 0.7678 - loss: 0.6864 - val accur
          acy: 0.5319 - val loss: 6139782.5000
          Epoch 3/10
          25/25 -
                                   - 2631s 110s/step - accuracy: 0.8477 - loss: 0.5850 - val_a
          ccuracy: 0.4397 - val loss: 7428.7617
          Epoch 4/10
          25/25 -
                                   - 43s 2s/step - accuracy: 0.9180 - loss: 0.2681 - val_accur
          acy: 0.1773 - val_loss: 1.4149
          Epoch 5/10
          25/25 -
                                    - 44s 2s/step - accuracy: 0.9460 - loss: 0.1602 - val accur
          acy: 0.1348 - val loss: 2.8440
          Epoch 6/10
          25/25 -
                                    - 45s 2s/step - accuracy: 0.9716 - loss: 0.0972 - val_accur
          acy: 0.1348 - val_loss: 4.9927
          Epoch 7/10
                               ——— 44s 2s/step - accuracy: 0.9862 - loss: 0.0400 - val_accur
          25/25 -
          acy: 0.1348 - val_loss: 7.1636
          Epoch 8/10
          25/25 -
                                    - 45s 2s/step - accuracy: 0.9698 - loss: 0.1068 - val accur
          acy: 0.1348 - val_loss: 12.4321
          Epoch 9/10
                                   - 44s 2s/step - accuracy: 0.9766 - loss: 0.0857 - val accur
          25/25 -
          acy: 0.1348 - val loss: 16.6639
          Epoch 10/10
          25/25 -
                                    - 44s 2s/step - accuracy: 0.9778 - loss: 0.0830 - val_accur
          acy: 0.1560 - val_loss: 8.0376
Out[64]: <keras.src.callbacks.history.History at 0x13e01b74850>
In [137]: | img=r"C:\Users\HP\Downloads\banana\BananaLSD\OriginalSet\sigatoka\51.jpeg"
In [138]: |img1=cv.imread(img)
In [139]: | img2=cv.resize(img1,(130,130))
```

In [140]: | img3=img2.reshape(1,130,130,3)

```
In [141]: plt.imshow(img1)
```

Out[141]: <matplotlib.image.AxesImage at 0x13ef0a3e250>



```
In [143]: np.argmax(pred3)
```

Out[143]: 3

In []:

InceptionV3

```
In [72]: from keras.applications import InceptionV3
model = InceptionV3(weights='imagenet')
```

```
In [73]: from keras.applications.inception_v3 import preprocess_input
```

```
In [74]: inception_v3=InceptionV3(input_shape=(130,130,3),weights='imagenet',include_top=False
```

In [81]: model.fit(F,T,epochs=20,validation_split=0.15)

```
Epoch 1/20
               50s 833ms/step - accuracy: 0.5034 - loss: 1.3270 - val_ac
25/25 ———
curacy: 0.3546 - val loss: 1.2859
Epoch 2/20
                 19s 747ms/step - accuracy: 0.8105 - loss: 0.5382 - val_ac
25/25 ———
curacy: 0.1560 - val loss: 1.3928
Epoch 3/20
25/25 -
                      —— 19s 740ms/step - accuracy: 0.9198 - loss: 0.2880 - val ac
curacy: 0.1844 - val_loss: 498.3741
Epoch 4/20
                       - 18s 739ms/step - accuracy: 0.8603 - loss: 0.3318 - val ac
25/25 -
curacy: 0.3546 - val_loss: 1.3311
Epoch 5/20
                19s 742ms/step - accuracy: 0.8600 - loss: 0.3101 - val_ac
25/25 ——
curacy: 0.4539 - val loss: 2.6314
Epoch 6/20
25/25 19s 749ms/step - accuracy: 0.9147 - loss: 0.2146 - val_ac
curacy: 0.4894 - val_loss: 6.1162
Epoch 7/20
                  19s 779ms/step - accuracy: 0.9025 - loss: 0.2625 - val_ac
25/25 -
curacy: 0.2695 - val_loss: 2.4585
Epoch 8/20
                20s 800ms/step - accuracy: 0.8657 - loss: 0.4041 - val_ac
25/25 -
curacy: 0.6667 - val loss: 2.3900
Epoch 9/20
               20s 809ms/step - accuracy: 0.8861 - loss: 0.4321 - val_ac
25/25 ———
curacy: 0.6099 - val loss: 4.6354
Epoch 10/20
                 ———— 21s 831ms/step - accuracy: 0.9250 - loss: 0.2590 - val ac
25/25 -
curacy: 0.6950 - val loss: 2.5515
Epoch 11/20
                 20s 788ms/step - accuracy: 0.9447 - loss: 0.2114 - val_ac
25/25 -
curacy: 0.6454 - val_loss: 1.3842
Epoch 12/20
                     21s 823ms/step - accuracy: 0.9202 - loss: 0.3020 - val ac
25/25 -
curacy: 0.6241 - val loss: 3.7017
Epoch 13/20
               21s 832ms/step - accuracy: 0.8976 - loss: 0.3499 - val_ac
25/25 ———
curacy: 0.5319 - val loss: 1.3736
Epoch 14/20
                 ———— 20s 807ms/step - accuracy: 0.9120 - loss: 0.2321 - val ac
curacy: 0.5035 - val_loss: 1.3500
Epoch 15/20
25/25 —
                    20s 802ms/step - accuracy: 0.9249 - loss: 0.2016 - val ac
curacy: 0.6028 - val_loss: 1.4164
Epoch 16/20
25/25 -
                       - 21s 830ms/step - accuracy: 0.9541 - loss: 0.1866 - val_ac
curacy: 0.7092 - val loss: 1.3011
Epoch 17/20
            21s 839ms/step - accuracy: 0.9439 - loss: 0.1615 - val_ac
curacy: 0.5319 - val loss: 1.8429
Epoch 18/20
                 20s 789ms/step - accuracy: 0.9240 - loss: 0.1888 - val ac
curacy: 0.2270 - val loss: 7.0645
Epoch 19/20
                     20s 783ms/step - accuracy: 0.9466 - loss: 0.1466 - val_ac
25/25 -
curacy: 0.6809 - val_loss: 1.7003
Epoch 20/20
```

```
- 19s 775ms/step - accuracy: 0.8865 - loss: 0.3664 - val_ac
          curacy: 0.5319 - val_loss: 9.9839
Out[81]: <keras.src.callbacks.history.History at 0x13e046dc850>
In [144]:
          img=r"C:\Users\HP\Downloads\banana\BananaLSD\OriginalSet\sigatoka\51.jpeg"
In [145]:
          img1=cv.imread(img)
In [146]:
          img2=cv.resize(img1,(130,130))
          img3=img2.reshape(1,130,130,3)
In [147]:
In [148]: plt.imshow(img1)
Out[148]: <matplotlib.image.AxesImage at 0x13ef0abae50>
              0
            25
             50
            75
            100
            125
            150
```



150

200

175

200 -

0

50

100