

Vegetable diseases prediction

```
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(r"C:\Users\Praveen\Desktop\FILES\data_for_ibm\Fertilizers_Recommendation_System_For_Disease_Prediction\Dataset Plant Disease\Veg-dataset\Veg-dataset\train_set", target_size=(128,128),  
                                         class_mode='categorical', batch_size=24)
```

Found 11385 images belonging to 9 classes.

```
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\Veg-dataset\Veg-dataset\test_set", target_size=(128,128),  
                                       class_mode='categorical', batch_size=24)
```

Found 3416 images belonging to 9 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

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

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

model.add(Dense(9,activation='softmax'))

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

len(x_train)

475

1238/2

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

475/475 [=====] - 237s 498ms/step - loss: 2.1787 - accuracy: 0.1331 - val_loss: 2.1362 - val_accuracy: 0.1953

Epoch 2/10

475/475 [=====] - 224s 470ms/step - loss: 2.1077 - accuracy: 0.1868 - val_loss: 2.1022 - val_accuracy: 0.1953

Epoch 3/10

475/475 [=====] - 242s 509ms/step - loss: 2.0872 - accuracy: 0.1868 - val_loss: 2.0911 - val_accuracy: 0.1953

Epoch 4/10

475/475 [=====] - 244s 514ms/step - loss: 2.0795 - accuracy: 0.1868 - val_loss: 2.0859 - val_accuracy: 0.1953

Epoch 5/10

475/475 [=====] - 249s 525ms/step - loss: 2.0761 - accuracy: 0.1868 - val_loss: 2.0846 - val_accuracy: 0.1953

Epoch 6/10

```

475/475 [=====] - 249s 525ms/step - loss: 2.0745 - accuracy:
0.1868 - val_loss: 2.0837 - val_accuracy: 0.1953
Epoch 7/10
475/475 [=====] - 250s 526ms/step - loss: 2.0738 - accuracy:
0.1868 - val_loss: 2.0830 - val_accuracy: 0.1953
Epoch 8/10
475/475 [=====] - 248s 521ms/step - loss: 2.0735 - accuracy:
0.1868 - val_loss: 2.0842 - val_accuracy: 0.1953
Epoch 9/10
475/475 [=====] - 221s 466ms/step - loss: 2.0734 - accuracy:
0.1868 - val_loss: 2.0844 - val_accuracy: 0.1953
Epoch 10/10
475/475 [=====] - 221s 465ms/step - loss: 2.0734 - accuracy:
0.1868 - val_loss: 2.0836 - val_accuracy: 0.1953

```

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

I

```

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

```

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

```

img=image.load_img(r"C:\Users\Praveen\Desktop\FILES\data_for_ibm\Fertilizers_Recommend
ation_System_For_Disease_Prediction\Dataset Plant Disease\Veg-dataset\Veg-
dataset\test_set\Potato___Early_blight\b817817e-a6b1-4123-88e7-
db98b453ce17___RS_Early.B 6880.jpg")

```

```
img
```

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

I

```

img=image.load_img(r"C:\Users\Praveen\Desktop\FILES\data_for_ibm\Fertilizers_Recommend
ation_System_For_Disease_Prediction\Dataset Plant Disease\Veg-dataset\Veg-
dataset\test_set\Potato___Early_blight\b817817e-a6b1-4123-88e7-
db98b453ce17___RS_Early.B 6880.jpg",target_size=(128,128))
img

```

```
x=image.img_to_array([[[[135., 131., 145.]
```

[134., 130., 144.],
[133., 129., 143.],

...,
[166., 164., 178.],
[188., 186., 200.],
[213., 211., 225.]],

[[141., 137., 151.],
[139., 135., 149.],
[128., 124., 138.],

...,
[201., 199., 213.],
[157., 155., 169.],
[172., 170., 184.]],

[[136., 132., 146.],
[135., 131., 145.],
[141., 137., 151.],

...,
[166., 164., 178.],
[169., 167., 181.],
[166., 164., 178.]],

...,

[[163., 161., 175.],
[154., 152., 166.],
[160., 158., 172.],

...,
[203., 201., 214.],
[221., 219., 232.],
[207., 205., 218.]],

[[148., 146., 160.],
[165., 163., 177.],
[152., 150., 164.],

...,
[176., 174., 187.],
[192., 190., 203.],
[189., 187., 200.]],

[[162., 160., 174.],
[155., 153., 167.],
[141., 139., 153.],

...,
[180., 178., 191.],

```
[190., 188., 201.],  
[191., 189., 202.]]], dtype=float32)
```

In [47]:

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

In [48]:

x

Out[48]:

```
array([[[[135., 131., 145.],  
        [134., 130., 144.],  
        [133., 129., 143.],  
        ...,  
        [166., 164., 178.],  
        [188., 186., 200.],  
        [213., 211., 225.]],  
       [[141., 137., 151.],  
        [139., 135., 149.],  
        [128., 124., 138.],  
        ...,  
        [201., 199., 213.],  
        [157., 155., 169.],  
        [172., 170., 184.]],  
       [[136., 132., 146.],  
        [135., 131., 145.],  
        [141., 137., 151.],  
        ...,  
        [166., 164., 178.],  
        [169., 167., 181.],  
        [166., 164., 178.]],  
       ...,  
       [[163., 161., 175.],  
        [154., 152., 166.],  
        [160., 158., 172.],  
        ...,  
        [203., 201., 214.],  
        [221., 219., 232.],  
        [207., 205., 218.]],  
       [[148., 146., 160.],  
        [165., 163., 177.],  
        [152., 150., 164.],  
        ...,  
        [176., 174., 187.],
```

```
[192., 190., 203.],  
[189., 187., 200.]],
```

```
[[162., 160., 174.],  
 [155., 153., 167.],  
 [141., 139., 153.],  
 ...,  
 [180., 178., 191.],  
 [190., 188., 201.],  
 [191., 189., 202.]]], dtype=float32)
```

In [49]:

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

```
1/1 [=====] - 0s 89ms/step
```

In [50]:

```
x_train.class_indices
```

Out[50]:

```
{'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 [51]:

```
index=['Pepper,_bell___Bacterial_spot','Pepper,_bell___healthy','Potato___Early_blight','Potato___Late_blight','Potato___healthy','Tomato___Bacterial_spot','Tomato___Late_blight','Tomato___Leaf_Mold','Tomato___Septoria_leaf_spot']
```

In [52]:

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

Out[52]:

```
'Tomato___Bacterial_spot'
```

In [53]:

```
img=image.load_img(r"C:\Users\Praveen\Desktop\FILES\data_for_ibm\Fertilizers_Recommendation_System_For_Disease_Prediction\Dataset Plant Disease\Veg-dataset\Veg-dataset\test_set\Potato___Early_blight\b817817e-a6b1-4123-88e7-db98b453ce17___RS_Early.B 6880.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___Early_blight','Potato___Late_blight','Potato___healthy','Tomato___Bacterial_spot','Tomato___Leaf_Mold','Tomato___Septoria_leaf_spot']
```

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