

```
In [2]: from tensorflow.keras.preprocessing.image import ImageDataGenerator
```

```
In [3]: train_datagen=ImageDataGenerator(rescale=1./255, zoom_range=0.2, horizontal_flip=True, vertical_flip=False)
```

```
In [4]: test_datagen=ImageDataGenerator(rescale=1./255)
```

```
In [5]: x_train=train_datagen.flow_from_directory(r"C:\Users\maris_q3mm6nk\Desktop\FILES\data_for_ibm\Fertilizers_Recommendation_System_For_Disease_Prediction",
class_mode='categorical', batch_size=24)
```

Found 11385 images belonging to 9 classes.

```
In [6]: x_test=test_datagen.flow_from_directory(r"C:\Users\maris_q3mm6nk\Desktop\FILES\data_for_ibm\Fertilizers_Recommendation_System_For_Disease_Prediction",
class_mode='categorical', batch_size=24)
```

Found 3416 images belonging to 9 classes.

```
In [7]: from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Convolution2D, MaxPooling2D, Flatten
```

```
In [8]: model=Sequential()
```

```
In [9]: model.add(Convolution2D(32,(3,3),input_shape=(128,128,3),activation='relu'))
```

```
In [10]: model.add(MaxPooling2D(pool_size=(2,2)))
```

```
In [11]: model.add(Flatten())
```

```
In [12]: model.summary()
```

Model: "sequential"

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		

```
In [13]: model.add(Dense(300,activation='relu'))
         model.add(Dense(150,activation='relu'))
```

```
In [20]: model.add(Dense(9,activation='softmax'))
```

```
In [21]: model.compile(loss='categorical_crossentropy',optimizer='adam',metrics=['accuracy'])
```

```
In [22]: len(x_train)
```

```
Out[22]: 475
```

```
In [23]: 1238/24
```

```
Out[23]: 51.583333333333336
```

```
In [24]: 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 1/10
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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

```

Out[24]:

```
In [25]: model.save('vegetabledata.h5')
```

```
In [26]: import numpy as np
from tensorflow.keras.models import load_model
from tensorflow.keras.preprocessing import image
```

```
In [27]: model=load_model('vegetabledata.h5')
```

```
In [33]: img=image.load_img(r"C:\Users\maris_q3mm6nk\Desktop\FILES\data_for_ibm\Fertilizers_Recommendation_System_For_Disease_Prediction\Dataset Plant Diseases")
```

```
In [34]: img
```

Out[34]:



In [35]: `x=image.img_to_array(img)`

In [44]: `img=image.load_img(r"C:\Users\maris_q3mm6nk\Desktop\FILES\data_for_ibm\Fertilizers_Recommendation_System_For_Disease_Prediction\Dataset Plant Diseases\img")`

Out[44]:



In [45]: `x=image.img_to_array(img)`

In [46]: `x`

```
Out[46]: 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_
```

```
In [52]: index[y[0]]
```

```
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_
```

```
In [52]: index[y[0]]
```

```
Out[52]: 'Tomato__Bacterial_spot'
```

```
In [53]: img=image.load_img(r"C:\Users\maris_q3mm6nk\Desktop\FILES\data_for_ibm\Fertilizers_Recommendation_System_For_Disease_Prediction\Dataset Plant Diseases  
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_  
index[y[0]]
```

```
1/1 [=====] - 0s 38ms/step
```

```
Out[53]: 'Tomato__Bacterial_spot'
```

```
In [ ]:
```