

Import necessary libraries

```
In [9]: import warnings
warnings.filterwarnings("ignore", category=DeprecationWarning)

import tensorflow as tf
import numpy as np
from tensorflow import keras
import matplotlib.pyplot as plt
import random
import os
import itertools
import datetime
from tensorflow.keras.layers.experimental.preprocessing import Rescaling
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from sklearn.model_selection import train_test_split
from tensorflow.keras import layers
from sklearn.metrics import precision_score, accuracy_score, recall_score, confusion_mat
```

Step 1: Data Augmentation

Define directories for dataset

```
In [7]: dataset_dir = 'Fruit And Vegetable Diseases Dataset'
```

Balance the dataset by oversampling

```
In [8]: def oversample_dataset(directory, target_size):
    datagen = ImageDataGenerator(
        rotation_range=20,
        width_shift_range=0.2,
        height_shift_range=0.2,
        shear_range=0.2,
        zoom_range=0.2,
        horizontal_flip=True,
        fill_mode='nearest'
    )

    class_folders = os.listdir(directory)

    for folder in class_folders:
        path = os.path.join(directory, folder)
        images = [os.path.join(path, img) for img in os.listdir(path)]

        # Check if oversampling is needed for the current class
        if len(images) < target_size:
            samples_to_add = target_size - len(images)
            if samples_to_add > 0:
                # Apply data augmentation for oversampling
                augmentation_gen = datagen.flow_from_directory(
                    directory=directory,
                    classes=[folder],
                    target_size=(224, 224),
                    batch_size=samples_to_add,
                    class_mode='categorical'
                )

            num_generated_images = 0
```

```

while num_generated_images < samples_to_add:
    batch = augmentation_gen.next()
    num_batch_images = batch[0].shape[0]
    for i in range(num_batch_images):
        if num_generated_images >= samples_to_add:
            break

    fruit_name = folder

    image = batch[0][i].squeeze()
    image_filename = f'{fruit_name}_augmented_{num_generated_images}'
    tf.keras.preprocessing.image.save_img(os.path.join(path, image_filename), image)
    images.append(os.path.join(path, image_filename))
    num_generated_images += 1

```

Balance and oversample the dataset

```

In [12]: target_size = 2000
oversample_dataset(dataset_dir, target_size)

```

```

Found 1641 images belonging to 1 classes.
Found 611 images belonging to 1 classes.
Found 591 images belonging to 1 classes.
Found 579 images belonging to 1 classes.
Found 619 images belonging to 1 classes.
Found 608 images belonging to 1 classes.
Found 593 images belonging to 1 classes.
Found 200 images belonging to 1 classes.
Found 200 images belonging to 1 classes.
Found 200 images belonging to 1 classes.
Found 200 images belonging to 1 classes.
Found 200 images belonging to 1 classes.
Found 200 images belonging to 1 classes.
Found 1813 images belonging to 1 classes.
Found 200 images belonging to 1 classes.
Found 200 images belonging to 1 classes.
Found 614 images belonging to 1 classes.

```

```

C:\Users\PMLS\anaconda3\Lib\site-packages\PIL\Image.py:970: UserWarning: Palette images with Transparency expressed in bytes should be converted to RGBA images
  warnings.warn(

```

```

Found 584 images belonging to 1 classes.
Found 1603 images belonging to 1 classes.
Found 1596 images belonging to 1 classes.
Found 604 images belonging to 1 classes.
Found 595 images belonging to 1 classes.

```

Seperate Train and Valid Dataset

```

In [13]: import os
import shutil
from sklearn.model_selection import train_test_split

def split_data(input_folder, output_folder, split_ratio):
    # List all subdirectories in the input folder
    subdirectories = [f.path for f in os.scandir(input_folder) if f.is_dir()]

    # Iterate through each subdirectory
    for subdirectory in subdirectories:
        # Get the class/category name from the subdirectory path
        class_name = os.path.basename(subdirectory)

        # List all files in the current subdirectory
        files = [f.path for f in os.scandir(subdirectory) if f.is_file()]

```

```

# Split files into training and testing sets
train_files, test_files = train_test_split(files, test_size=split_ratio, random_

# Create output folders for training and testing sets
train_output_folder = os.path.join(output_folder, 'train', class_name)
test_output_folder = os.path.join(output_folder, 'valid', class_name)
os.makedirs(train_output_folder, exist_ok=True)
os.makedirs(test_output_folder, exist_ok=True)

# Copy training files to the training output folder
for train_file in train_files:
    shutil.copy(train_file, train_output_folder)

# Copy testing files to the testing output folder
for test_file in test_files:
    shutil.copy(test_file, test_output_folder)

# Path to the output directory where the train and valid folders will be created
output_dir = 'Fruit And Veg Diseases Dataset'
split_ratio = 0.2 # Adjust the split ratio as needed

split_data(dataset_dir, output_dir, split_ratio)

```

Step 2: Data Preparation

Define directories for training and testing data

```

In [1]: train_dir = 'Fruit And Veg Diseases Dataset/train'
        test_dir = 'Fruit And Veg Diseases Dataset/valid'

```

```

In [4]: from PIL import Image
        import os

        def filter_images_by_format(directory, allowed_formats):
            for root, dirs, files in os.walk(directory):
                for file in files:
                    file_path = os.path.join(root, file)
                    try:
                        # Open the image to check its format
                        with Image.open(file_path) as img:
                            img_format = img.format.upper()

                        # Check if the format is not in the allowed_formats list
                        if img_format not in allowed_formats:
                            print(f"Deleting {file_path} (unsupported format: {img_format})")
                            os.remove(file_path)

                    except Exception as e:
                        print(f"Error processing {file_path}: {e}")

        # Specify the allowed image formats
        allowed_formats = ['JPEG', 'PNG', 'JPG', 'GIF', 'BMP']

        # Apply the filter to the training directory
        filter_images_by_format(train_dir, allowed_formats)

        # Apply the filter to the testing directory
        filter_images_by_format(test_dir, allowed_formats)

```

Deleting Fruit And Veg Diseases Dataset/train\Potato__Rotten\rottenPotato (1).webp (unsupported format: WEBP)

Deleting Fruit And Veg Diseases Dataset/train\Tomato__Rotten\rottenTomato (1).webp (unsu

```
pported format: WEBP)
Deleting Fruit And Veg Diseases Dataset/valid\Banana__Healthy\freshBanana (1).webp (unsu
pported format: WEBP)
Deleting Fruit And Veg Diseases Dataset/valid\Banana__Rotten\rottenBanana (1).webp (unsu
pported format: WEBP)
Deleting Fruit And Veg Diseases Dataset/valid\Carrot__Healthy\freshCarrot (415).jpg (uns
upported format: WEBP)
```

```
In [10]: train_dir = 'Fruit And Veg Diseases Dataset/train'
        test_dir = 'Fruit And Veg Diseases Dataset/valid'
```

Create image datasets for training and testing

```
In [11]: train_data = keras.utils.image_dataset_from_directory(train_dir,
                                                             image_size=(224, 224),
                                                             label_mode='categorical',
                                                             batch_size=32)
```

Found 46924 files belonging to 28 classes.

```
In [12]: test_data = keras.utils.image_dataset_from_directory(test_dir,
                                                             image_size=(224, 224),
                                                             label_mode='categorical',
                                                             batch_size=32)
```

Found 11734 files belonging to 28 classes.

Define class names based on the directory structure

```
In [13]: class_names = train_data.class_names
        class_names
```

```
Out[13]: ['Apple__Healthy',
          'Apple__Rotten',
          'Banana__Healthy',
          'Banana__Rotten',
          'Bellpepper__Healthy',
          'Bellpepper__Rotten',
          'Carrot__Healthy',
          'Carrot__Rotten',
          'Cucumber__Healthy',
          'Cucumber__Rotten',
          'Grape__Healthy',
          'Grape__Rotten',
          'Guava__Healthy',
          'Guava__Rotten',
          'Jujube__Healthy',
          'Jujube__Rotten',
          'Mango__Healthy',
          'Mango__Rotten',
          'Orange__Healthy',
          'Orange__Rotten',
          'Pomegranate__Healthy',
          'Pomegranate__Rotten',
          'Potato__Healthy',
          'Potato__Rotten',
          'Strawberry__Healthy',
          'Strawberry__Rotten',
          'Tomato__Healthy',
          'Tomato__Rotten']
```

Step 2: Model Creation

Define the input image shape

```
In [14]: image_shape = (224, 224, 3)
```

Create a base model (EfficientNetB0) for feature extraction

```
In [15]: base_model = tf.keras.applications.EfficientNetB0(include_top=False, weights='imagenet')
base_model.trainable = False
```

WARNING:tensorflow:From C:\Users\PMLS\anaconda3\Lib\site-packages\keras\src\backend.py:1398: The name tf.executing_eagerly_outside_functions is deprecated. Please use tf.compat.v1.executing_eagerly_outside_functions instead.

WARNING:tensorflow:From C:\Users\PMLS\anaconda3\Lib\site-packages\keras\src\layers\normalization\batch_normalization.py:979: The name tf.nn.fused_batch_norm is deprecated. Please use tf.compat.v1.nn.fused_batch_norm instead.

Downloading data from https://storage.googleapis.com/keras-applications/efficientnetb0_n
otop.h5
16705208/16705208 [=====] - 19s 1us/step

Create the main model by adding layers on top of the base model

```
In [16]: inputs = layers.Input(shape=image_shape, name='input_layer')
x = base_model(inputs, training=False)
x = layers.GlobalAveragePooling2D(name='GlobalAveragePooling2D_layer')(x)
outputs = layers.Dense(len(class_names), activation='softmax', name='output_layer')(x)
feature_model = tf.keras.Model(inputs, outputs, name='Fruit_Vegetable_Diseases_Detection')
```

Set some layers in the base model as trainable

```
In [17]: base_model.trainable = True
for layer in base_model.layers[:-20]:
    layer.trainable = False
```

Compile the model

```
In [18]: feature_model.compile(
    loss='categorical_crossentropy',
    optimizer=tf.keras.optimizers.Adam(learning_rate=0.0001),
    metrics=['accuracy']
)
```

```
In [19]: base_model.summary()
```

Model: "efficientnetb0"

Layer (type)	Output Shape	Param #	Connected to
=====			
input_1 (InputLayer)	[(None, None, None, 3)]	0	[]
rescaling (Rescaling)	(None, None, None, 3)	0	['input_1[0][0]']
normalization (Normalizati	(None, None, None, 3)	7	['rescaling[0][0]']

on)				
rescaling_1 (Rescaling)	(None, None, None, 3)	0	['normalization[0][0]']	
stem_conv_pad (ZeroPadding2D)	(None, None, None, 3)	0	['rescaling_1[0][0]']	
stem_conv (Conv2D)	(None, None, None, 32)	864	['stem_conv_pad[0][0]']	
stem_bn (BatchNormalization)	(None, None, None, 32)	128	['stem_conv[0][0]']	
stem_activation (Activation)	(None, None, None, 32)	0	['stem_bn[0][0]']	
block1a_dwconv (DepthwiseConv2D)	(None, None, None, 32)	288	['stem_activation[0][0]']	
block1a_bn (BatchNormalization)	(None, None, None, 32)	128	['block1a_dwconv[0][0]']	
block1a_activation (Activation)	(None, None, None, 32)	0	['block1a_bn[0][0]']	
block1a_se_squeeze (GlobalAveragePooling2D)	(None, 32)	0	['block1a_activation[0][0]']	
block1a_se_reshape (Reshape)	(None, 1, 1, 32)	0	['block1a_se_squeeze[0][0]']	
block1a_se_reduce (Conv2D)	(None, 1, 1, 8)	264	['block1a_se_reshape[0][0]']	
block1a_se_expand (Conv2D)	(None, 1, 1, 32)	288	['block1a_se_reduce[0][0]']	

block1a_se_excite (Multipl [0][0]', y) [0][0]']	(None, None, None, 32)	0	['block1a_activation 'block1a_se_expand
block1a_project_conv (Conv [0][0]'] 2D)	(None, None, None, 16)	512	['block1a_se_excite
block1a_project_bn (BatchN nv[0][0]'] ormalization)	(None, None, None, 16)	64	['block1a_project_co
block2a_expand_conv (Conv2 [0][0]'] D)	(None, None, None, 96)	1536	['block1a_project_bn
block2a_expand_bn (BatchNo v[0][0]'] rmalization)	(None, None, None, 96)	384	['block2a_expand_con
block2a_expand_activation [0][0]'] (Activation)	(None, None, None, 96)	0	['block2a_expand_bn
block2a_dwconv_pad (ZeroPa ivation[0] dding2D)	(None, None, None, 96)	0	['block2a_expand_act [0]']
block2a_dwconv (DepthwiseC [0][0]'] onv2D)	(None, None, None, 96)	864	['block2a_dwconv_pad
block2a_bn (BatchNormaliza [0]'] tion)	(None, None, None, 96)	384	['block2a_dwconv[0]
block2a_activation (Activa tion)	(None, None, None, 96)	0	['block2a_bn[0][0]']
block2a_se_squeeze (Global [0][0]'] AveragePooling2D)	(None, 96)	0	['block2a_activation

block2a_se_reshape (Reshape) [0][0]'] e)	(None, 1, 1, 96)	0	['block2a_se_squeeze
block2a_se_reduce (Conv2D) [0][0]']	(None, 1, 1, 4)	388	['block2a_se_reshape
block2a_se_expand (Conv2D) [0][0]']	(None, 1, 1, 96)	480	['block2a_se_reduce
block2a_se_excite (Multipl [0][0]', y) [0][0]']	(None, None, None, 96)	0	['block2a_activation 'block2a_se_expand
block2a_project_conv (Conv [0][0]'] 2D)	(None, None, None, 24)	2304	['block2a_se_excite
block2a_project_bn (BatchN nv[0][0]'] ormalization)	(None, None, None, 24)	96	['block2a_project_co
block2b_expand_conv (Conv2 [0][0]'] D)	(None, None, None, 144)	3456	['block2a_project_bn
block2b_expand_bn (BatchNo v[0][0]'] rmalization)	(None, None, None, 144)	576	['block2b_expand_con
block2b_expand_activation [0][0]'] (Activation)	(None, None, None, 144)	0	['block2b_expand_bn
block2b_dwconv (DepthwiseC ivation[0] onv2D)	(None, None, None, 144)	1296	['block2b_expand_act [0]']
block2b_bn (BatchNormaliza [0]'] tion)	(None, None, None, 144)	576	['block2b_dwconv[0]
block2b_activation (Activa	(None, None, None, 144)	0	['block2b_bn[0][0]']

tion)				
block2b_se_squeeze (Global AveragePooling2D)	(None, 144)	0	['block2b_activation[0][0]']	
block2b_se_reshape (Reshape)	(None, 1, 1, 144)	0	['block2b_se_squeeze[0][0]']	
block2b_se_reduce (Conv2D)	(None, 1, 1, 6)	870	['block2b_se_reshape[0][0]']	
block2b_se_expand (Conv2D)	(None, 1, 1, 144)	1008	['block2b_se_reduce[0][0]']	
block2b_se_excite (Multiply)	(None, None, None, 144)	0	['block2b_activation[0][0]', 'y' 'block2b_se_expand[0][0]']	
block2b_project_conv (Conv2D)	(None, None, None, 24)	3456	['block2b_se_excite[0][0]']	
block2b_project_bn (BatchNormalization)	(None, None, None, 24)	96	['block2b_project_conv[0][0]']	
block2b_drop (Dropout)	(None, None, None, 24)	0	['block2b_project_bn[0][0]']	
block2b_add (Add)	(None, None, None, 24)	0	['block2b_drop[0][0]', 'block2a_project_bn[0][0]']	
block3a_expand_conv (Conv2D)	(None, None, None, 144)	3456	['block2b_add[0][0]']	
block3a_expand_bn (BatchNormalization)	(None, None, None, 144)	576	['block3a_expand_conv[0][0]']	
block3a_expand_activation	(None, None, None, 144)	0	['block3a_expand_bn[0][0]']	

(Activation)			
block3a_dwconv_pad (ZeroPadding2D)	(None, None, None, 144)	0	['block3a_expand_activation[0][0]']
block3a_dwconv (DepthwiseConv2D)	(None, None, None, 144)	3600	['block3a_dwconv_pad[0][0]']
block3a_bn (BatchNormalization)	(None, None, None, 144)	576	['block3a_dwconv[0][0]']
block3a_activation (Activation)	(None, None, None, 144)	0	['block3a_bn[0][0]']
block3a_se_squeeze (GlobalAveragePooling2D)	(None, 144)	0	['block3a_activation[0][0]']
block3a_se_reshape (Reshape)	(None, 1, 1, 144)	0	['block3a_se_squeeze[0][0]']
block3a_se_reduce (Conv2D)	(None, 1, 1, 6)	870	['block3a_se_reshape[0][0]']
block3a_se_expand (Conv2D)	(None, 1, 1, 144)	1008	['block3a_se_reduce[0][0]']
block3a_se_excite (Multiply)	(None, None, None, 144)	0	['block3a_activation[0][0]', 'block3a_se_expand[0][0]']
block3a_project_conv (Conv2D)	(None, None, None, 40)	5760	['block3a_se_excite[0][0]']
block3a_project_bn (BatchNormalization)	(None, None, None, 40)	160	['block3a_project_conv[0][0]']

block3b_expand_conv (Conv2 [0][0]') D)	(None, None, None, 240)	9600	['block3a_project_bn
block3b_expand_bn (BatchNo v[0][0]') rmalization)	(None, None, None, 240)	960	['block3b_expand_con
block3b_expand_activation [0][0]') (Activation)	(None, None, None, 240)	0	['block3b_expand_bn
block3b_dwconv (DepthwiseC ivation[0] onv2D)	(None, None, None, 240)	6000	['block3b_expand_act [0]']
block3b_bn (BatchNormaliza [0]') tion)	(None, None, None, 240)	960	['block3b_dwconv[0]
block3b_activation (Activa tion)	(None, None, None, 240)	0	['block3b_bn[0][0]']
block3b_se_squeeze (Global [0][0]') AveragePooling2D)	(None, 240)	0	['block3b_activation
block3b_se_reshape (Reshap [0][0]') e)	(None, 1, 1, 240)	0	['block3b_se_squeeze
block3b_se_reduce (Conv2D) [0][0]')	(None, 1, 1, 10)	2410	['block3b_se_reshape
block3b_se_expand (Conv2D) [0][0]')	(None, 1, 1, 240)	2640	['block3b_se_reduce
block3b_se_excite (Multipl [0][0]', y) [0][0]')	(None, None, None, 240)	0	['block3b_activation 'block3b_se_expand
block3b_project_conv (Conv [0][0]') 2D)	(None, None, None, 40)	9600	['block3b_se_excite

block3b_project_bn (BatchNormal ization)	(None, None, None, 40)	160	['block3b_project_co nv[0][0]']
block3b_drop (Dropout)	(None, None, None, 40)	0	['block3b_project_bn [0][0]']
block3b_add (Add)	(None, None, None, 40)	0	['block3b_drop[0] [0]', 'block3a_project_bn [0][0]']
block4a_expand_conv (Conv2 D)	(None, None, None, 240)	9600	['block3b_add[0] [0]']
block4a_expand_bn (BatchNo rmalization)	(None, None, None, 240)	960	['block4a_expand_con v[0][0]']
block4a_expand_activation (Activation)	(None, None, None, 240)	0	['block4a_expand_bn [0][0]']
block4a_dwconv_pad (ZeroPa dding2D)	(None, None, None, 240)	0	['block4a_expand_act ivation[0] [0]']
block4a_dwconv (DepthwiseC onv2D)	(None, None, None, 240)	2160	['block4a_dwconv_pad [0][0]']
block4a_bn (BatchNormaliza tion)	(None, None, None, 240)	960	['block4a_dwconv[0] [0]']
block4a_activation (Activa tion)	(None, None, None, 240)	0	['block4a_bn[0][0]']
block4a_se_squeeze (Global AveragePooling2D)	(None, 240)	0	['block4a_activation [0][0]']

block4a_se_reshape (Reshape) [0][0]'] e)	(None, 1, 1, 240)	0	['block4a_se_squeeze
block4a_se_reduce (Conv2D) [0][0]']	(None, 1, 1, 10)	2410	['block4a_se_reshape
block4a_se_expand (Conv2D) [0][0]']	(None, 1, 1, 240)	2640	['block4a_se_reduce
block4a_se_excite (Multipl [0][0]', y) [0][0]']	(None, None, None, 240)	0	['block4a_activation 'block4a_se_expand
block4a_project_conv (Conv [0][0]'] 2D)	(None, None, None, 80)	19200	['block4a_se_excite
block4a_project_bn (BatchN nv[0][0]'] ormalization)	(None, None, None, 80)	320	['block4a_project_co
block4b_expand_conv (Conv2 [0][0]'] D)	(None, None, None, 480)	38400	['block4a_project_bn
block4b_expand_bn (BatchNo v[0][0]'] rmalization)	(None, None, None, 480)	1920	['block4b_expand_con
block4b_expand_activation [0][0]'] (Activation)	(None, None, None, 480)	0	['block4b_expand_bn
block4b_dwconv (DepthwiseC ivation[0] onv2D)	(None, None, None, 480)	4320	['block4b_expand_act [0]']
block4b_bn (BatchNormaliza [0]'] tion)	(None, None, None, 480)	1920	['block4b_dwconv[0]
block4b_activation (Activa tion)	(None, None, None, 480)	0	['block4b_bn[0][0]']

block4b_se_squeeze (Global [0][0]'] AveragePooling2D)	(None, 480)	0	['block4b_activation
block4b_se_reshape (Reshap [0][0]'] e)	(None, 1, 1, 480)	0	['block4b_se_squeeze
block4b_se_reduce (Conv2D) [0][0]']	(None, 1, 1, 20)	9620	['block4b_se_reshape
block4b_se_expand (Conv2D) [0][0]']	(None, 1, 1, 480)	10080	['block4b_se_reduce
block4b_se_excite (Multipl [0][0]', y) [0][0]']	(None, None, None, 480)	0	['block4b_activation 'block4b_se_expand
block4b_project_conv (Conv [0][0]'] 2D)	(None, None, None, 80)	38400	['block4b_se_excite
block4b_project_bn (BatchN nv[0][0]'] ormalization)	(None, None, None, 80)	320	['block4b_project_co
block4b_drop (Dropout) [0][0]']	(None, None, None, 80)	0	['block4b_project_bn
block4b_add (Add) [0]', [0][0]']	(None, None, None, 80)	0	['block4b_drop[0] 'block4a_project_bn
block4c_expand_conv (Conv2 [0]'] D)	(None, None, None, 480)	38400	['block4b_add[0]
block4c_expand_bn (BatchNo v[0][0]'] rmalization)	(None, None, None, 480)	1920	['block4c_expand_con
block4c_expand_activation [0][0]'] (Activation)	(None, None, None, 480)	0	['block4c_expand_bn

block4c_dwconv (DepthwiseConv2D)	(None, None, None, 480)	4320	['block4c_expand_activation[0][0]']
block4c_bn (BatchNormalization)	(None, None, None, 480)	1920	['block4c_dwconv[0][0]']
block4c_activation (Activation)	(None, None, None, 480)	0	['block4c_bn[0][0]']
block4c_se_squeeze (GlobalAveragePooling2D)	(None, 480)	0	['block4c_activation[0][0]']
block4c_se_reshape (Reshape)	(None, 1, 1, 480)	0	['block4c_se_squeeze[0][0]']
block4c_se_reduce (Conv2D)	(None, 1, 1, 20)	9620	['block4c_se_reshape[0][0]']
block4c_se_expand (Conv2D)	(None, 1, 1, 480)	10080	['block4c_se_reduce[0][0]']
block4c_se_excite (Multiply)	(None, None, None, 480)	0	['block4c_activation[0][0]', 'block4c_se_expand[0][0]']
block4c_project_conv (Conv2D)	(None, None, None, 80)	38400	['block4c_se_excite[0][0]']
block4c_project_bn (BatchNormalization)	(None, None, None, 80)	320	['block4c_project_conv[0][0]']
block4c_drop (Dropout)	(None, None, None, 80)	0	['block4c_project_bn[0][0]']
block4c_add (Add)	(None, None, None, 80)	0	['block4c_drop[0][0]', 'block4b_add[0][0]']

block5a_expand_conv (Conv2D)	(None, None, None, 480)	38400	['block4c_add[0]
block5a_expand_bn (BatchNormalization)	(None, None, None, 480)	1920	['block5a_expand_con
block5a_expand_activation (Activation)	(None, None, None, 480)	0	['block5a_expand_bn
block5a_dwconv (DepthwiseConv2D)	(None, None, None, 480)	12000	['block5a_expand_act
block5a_bn (BatchNormalization)	(None, None, None, 480)	1920	['block5a_dwconv[0]
block5a_activation (Activation)	(None, None, None, 480)	0	['block5a_bn[0][0]']
block5a_se_squeeze (GlobalAveragePooling2D)	(None, 480)	0	['block5a_activation
block5a_se_reshape (Reshape)	(None, 1, 1, 480)	0	['block5a_se_squeeze
block5a_se_reduce (Conv2D)	(None, 1, 1, 20)	9620	['block5a_se_reshape
block5a_se_expand (Conv2D)	(None, 1, 1, 480)	10080	['block5a_se_reduce
block5a_se_excite (Multiply)	(None, None, None, 480)	0	['block5a_activation
block5a_project_conv (Conv2D)	(None, None, None, 112)	53760	['block5a_se_excite

2D)

block5a_project_bn (BatchNormaliza tion[0][0])	(None, None, None, 112)	448	['block5a_project_co nv[0][0]']
block5b_expand_conv (Conv2D)	(None, None, None, 672)	75264	['block5a_project_bn [0][0]']
block5b_expand_bn (BatchNormaliza tion[0][0])	(None, None, None, 672)	2688	['block5b_expand_con v[0][0]']
block5b_expand_activation (Activation)	(None, None, None, 672)	0	['block5b_expand_bn [0][0]']
block5b_dwconv (DepthwiseConv2D)	(None, None, None, 672)	16800	['block5b_expand_act ivation[0] conv2D']
block5b_bn (BatchNormalization)	(None, None, None, 672)	2688	['block5b_dwconv[0] [0]']
block5b_activation (Activation)	(None, None, None, 672)	0	['block5b_bn[0][0]']
block5b_se_squeeze (GlobalAveragePooling2D)	(None, 672)	0	['block5b_activation [0][0]']
block5b_se_reshape (Reshape)	(None, 1, 1, 672)	0	['block5b_se_squeeze [0][0]']
block5b_se_reduce (Conv2D)	(None, 1, 1, 28)	18844	['block5b_se_reshape [0][0]']
block5b_se_expand (Conv2D)	(None, 1, 1, 672)	19488	['block5b_se_reduce [0][0]']

block5b_se_excite (Multipl [0][0]', y) [0][0]']	(None, None, None, 672)	0	['block5b_activation
block5b_project_conv (Conv [0][0]'] 2D)	(None, None, None, 112)	75264	['block5b_se_excite
block5b_project_bn (BatchN nv[0][0]'] ormalization)	(None, None, None, 112)	448	['block5b_project_co
block5b_drop (Dropout) [0][0]']	(None, None, None, 112)	0	['block5b_project_bn
block5b_add (Add) [0]', [0][0]']	(None, None, None, 112)	0	['block5b_drop[0] 'block5a_project_bn
block5c_expand_conv (Conv2 [0]'] D)	(None, None, None, 672)	75264	['block5b_add[0]
block5c_expand_bn (BatchNo v[0][0]'] rmalization)	(None, None, None, 672)	2688	['block5c_expand_con
block5c_expand_activation [0][0]'] (Activation)	(None, None, None, 672)	0	['block5c_expand_bn
block5c_dwconv (DepthwiseC ivation[0] onv2D)	(None, None, None, 672)	16800	['block5c_expand_act [0]']
block5c_bn (BatchNormaliza [0]'] tion)	(None, None, None, 672)	2688	['block5c_dwconv[0]
block5c_activation (Activa tion)	(None, None, None, 672)	0	['block5c_bn[0][0]']
block5c_se_squeeze (Global [0][0]']	(None, 672)	0	['block5c_activation

AveragePooling2D)

block5c_se_reshape (Reshape)	(None, 1, 1, 672)	0	['block5c_se_squeeze[0][0]']
block5c_se_reduce (Conv2D)	(None, 1, 1, 28)	18844	['block5c_se_reshape[0][0]']
block5c_se_expand (Conv2D)	(None, 1, 1, 672)	19488	['block5c_se_reduce[0][0]']
block5c_se_excite (Multiply)	(None, None, None, 672)	0	['block5c_activation[0][0]', 'block5c_se_expand[0][0]']
block5c_project_conv (Conv2D)	(None, None, None, 112)	75264	['block5c_se_excite[0][0]']
block5c_project_bn (Batch Normalization)	(None, None, None, 112)	448	['block5c_project_conv[0][0]']
block5c_drop (Dropout)	(None, None, None, 112)	0	['block5c_project_bn[0][0]']
block5c_add (Add)	(None, None, None, 112)	0	['block5c_drop[0][0]', 'block5b_add[0][0]']
block6a_expand_conv (Conv2D)	(None, None, None, 672)	75264	['block5c_add[0][0]']
block6a_expand_bn (Batch Normalization)	(None, None, None, 672)	2688	['block6a_expand_conv[0][0]']
block6a_expand_activation (Activation)	(None, None, None, 672)	0	['block6a_expand_bn[0][0]']
block6a_dwconv_pad (ZeroPadding2D)	(None, None, None, 672)	0	['block6a_expand_activation[0][0]']

dding2D)			[0]']
block6a_dwconv (DepthwiseConv2D)	(None, None, None, 672)	16800	['block6a_dwconv_pad[0][0]']
block6a_bn (BatchNormalization)	(None, None, None, 672)	2688	['block6a_dwconv[0][0]']
block6a_activation (Activation)	(None, None, None, 672)	0	['block6a_bn[0][0]']
block6a_se_squeeze (GlobalAveragePooling2D)	(None, 672)	0	['block6a_activation[0][0]']
block6a_se_reshape (Reshape)	(None, 1, 1, 672)	0	['block6a_se_squeeze[0][0]']
block6a_se_reduce (Conv2D)	(None, 1, 1, 28)	18844	['block6a_se_reshape[0][0]']
block6a_se_expand (Conv2D)	(None, 1, 1, 672)	19488	['block6a_se_reduce[0][0]']
block6a_se_excite (Multiply)	(None, None, None, 672)	0	['block6a_activation[0][0]', 'block6a_se_expand[0][0]']
block6a_project_conv (Conv2D)	(None, None, None, 192)	129024	['block6a_se_excite[0][0]']
block6a_project_bn (BatchNormalization)	(None, None, None, 192)	768	['block6a_project_conv[0][0]']
block6b_expand_conv (Conv2D)	(None, None, None, 1152)	221184	['block6a_project_bn[0][0]']

block6b_expand_bn (BatchNormalization)	(None, None, None, 1152)	4608	['block6b_expand_con
block6b_expand_activation (Activation)	(None, None, None, 1152)	0	['block6b_expand_bn
block6b_dwconv (DepthwiseConv2D)	(None, None, None, 1152)	28800	['block6b_expand_act
block6b_bn (BatchNormalization)	(None, None, None, 1152)	4608	['block6b_dwconv[0]
block6b_activation (Activation)	(None, None, None, 1152)	0	['block6b_bn[0][0]']
block6b_se_squeeze (GlobalAveragePooling2D)	(None, 1152)	0	['block6b_activation
block6b_se_reshape (Reshape)	(None, 1, 1, 1152)	0	['block6b_se_squeeze
block6b_se_reduce (Conv2D)	(None, 1, 1, 48)	55344	['block6b_se_reshape
block6b_se_expand (Conv2D)	(None, 1, 1, 1152)	56448	['block6b_se_reduce
block6b_se_excite (Multiply)	(None, None, None, 1152)	0	['block6b_activation
block6b_project_conv (Conv2D)	(None, None, None, 192)	221184	['block6b_se_expand
block6b_project_bn (BatchNormalization)	(None, None, None, 192)	768	['block6b_project_co

block6b_drop (Dropout) [0][0]']	(None, None, None, 192)	0	['block6b_project_bn
block6b_add (Add) [0]', [0][0]']	(None, None, None, 192)	0	['block6b_drop[0] 'block6a_project_bn
block6c_expand_conv (Conv2 [0]'] D)	(None, None, None, 1152)	221184	['block6b_add[0]
block6c_expand_bn (BatchNo v[0][0]'] rmalization)	(None, None, None, 1152)	4608	['block6c_expand_con
block6c_expand_activation [0][0]'] (Activation)	(None, None, None, 1152)	0	['block6c_expand_bn
block6c_dwconv (DepthwiseC ivation[0] onv2D)	(None, None, None, 1152)	28800	['block6c_expand_act [0]']
block6c_bn (BatchNormaliza [0]'] tion)	(None, None, None, 1152)	4608	['block6c_dwconv[0]
block6c_activation (Activa tion)	(None, None, None, 1152)	0	['block6c_bn[0][0]']
block6c_se_squeeze (Global [0][0]'] AveragePooling2D)	(None, 1152)	0	['block6c_activation
block6c_se_reshape (Reshap [0][0]'] e)	(None, 1, 1, 1152)	0	['block6c_se_squeeze
block6c_se_reduce (Conv2D) [0][0]']	(None, 1, 1, 48)	55344	['block6c_se_reshape
block6c_se_expand (Conv2D) [0][0]']	(None, 1, 1, 1152)	56448	['block6c_se_reduce

block6c_se_excite (Multipl [0][0]', y) [0][0]']	(None, None, None, 1152)	0	['block6c_activation 'block6c_se_expand
block6c_project_conv (Conv [0][0]'] 2D)	(None, None, None, 192)	221184	['block6c_se_excite
block6c_project_bn (BatchN nv[0][0]'] ormalization)	(None, None, None, 192)	768	['block6c_project_co
block6c_drop (Dropout) [0][0]']	(None, None, None, 192)	0	['block6c_project_bn
block6c_add (Add) [0]', [0]']	(None, None, None, 192)	0	['block6c_drop[0] 'block6b_add[0]
block6d_expand_conv (Conv2 [0]'] D)	(None, None, None, 1152)	221184	['block6c_add[0]
block6d_expand_bn (BatchNo v[0][0]'] rmalization)	(None, None, None, 1152)	4608	['block6d_expand_con
block6d_expand_activation [0][0]'] (Activation)	(None, None, None, 1152)	0	['block6d_expand_bn
block6d_dwconv (DepthwiseC ivation[0] onv2D)	(None, None, None, 1152)	28800	['block6d_expand_act [0]']
block6d_bn (BatchNormaliza [0]'] tion)	(None, None, None, 1152)	4608	['block6d_dwconv[0]
block6d_activation (Activa tion)	(None, None, None, 1152)	0	['block6d_bn[0][0]']

block6d_se_squeeze (Global [0][0]'] AveragePooling2D)	(None, 1152)	0	['block6d_activation
block6d_se_reshape (Reshap [0][0]'] e)	(None, 1, 1, 1152)	0	['block6d_se_squeeze
block6d_se_reduce (Conv2D) [0][0]']	(None, 1, 1, 48)	55344	['block6d_se_reshape
block6d_se_expand (Conv2D) [0][0]']	(None, 1, 1, 1152)	56448	['block6d_se_reduce
block6d_se_excite (Multipl [0][0]', y) [0][0]']	(None, None, None, 1152)	0	['block6d_activation 'block6d_se_expand
block6d_project_conv (Conv [0][0]'] 2D)	(None, None, None, 192)	221184	['block6d_se_excite
block6d_project_bn (BatchN nv[0][0]'] ormalization)	(None, None, None, 192)	768	['block6d_project_co
block6d_drop (Dropout) [0][0]']	(None, None, None, 192)	0	['block6d_project_bn
block6d_add (Add) [0]', [0]']	(None, None, None, 192)	0	['block6d_drop[0] 'block6c_add[0]
block7a_expand_conv (Conv2 [0]'] D)	(None, None, None, 1152)	221184	['block6d_add[0]
block7a_expand_bn (BatchNo v[0][0]'] rmalization)	(None, None, None, 1152)	4608	['block7a_expand_con
block7a_expand_activation [0][0]'] (Activation)	(None, None, None, 1152)	0	['block7a_expand_bn

block7a_dwconv (DepthwiseConv2D)	(None, None, None, 1152)	10368	['block7a_expand_activation[0][0]']
block7a_bn (BatchNormalization)	(None, None, None, 1152)	4608	['block7a_dwconv[0][0]']
block7a_activation (Activation)	(None, None, None, 1152)	0	['block7a_bn[0][0]']
block7a_se_squeeze (GlobalAveragePooling2D)	(None, 1152)	0	['block7a_activation[0][0]']
block7a_se_reshape (Reshape)	(None, 1, 1, 1152)	0	['block7a_se_squeeze[0][0]']
block7a_se_reduce (Conv2D)	(None, 1, 1, 48)	55344	['block7a_se_reshape[0][0]']
block7a_se_expand (Conv2D)	(None, 1, 1, 1152)	56448	['block7a_se_reduce[0][0]']
block7a_se_excite (Multiply)	(None, None, None, 1152)	0	['block7a_activation[0][0]', 'block7a_se_expand[0][0]']
block7a_project_conv (Conv2D)	(None, None, None, 320)	368640	['block7a_se_excite[0][0]']
block7a_project_bn (BatchNormalization)	(None, None, None, 320)	1280	['block7a_project_conv[0][0]']
top_conv (Conv2D)	(None, None, None, 1280)	409600	['block7a_project_bn[0][0]']
top_bn (BatchNormalization)	(None, None, None, 1280)	5120	['top_conv[0][0]']

```

top_activation (Activation (None, None, None, 1280) 0 ['top_bn[0][0]'])
)

=====
Total params: 4049571 (15.45 MB)
Trainable params: 1350960 (5.15 MB)
Non-trainable params: 2698611 (10.29 MB)

```

Step 3: Model Training

Create a function to set up TensorBoard logging

```

In [20]: def create_tensorboard_callback(dir_name, experiment_name):
        log_dir = dir_name + "/" + experiment_name + "/" + datetime.datetime.now().strftime(
            '%Y%m%d-%H%M%S')
        tensorboard_callback = tf.keras.callbacks.TensorBoard(
            log_dir=log_dir
        )
        print(f"Saving TensorBoard log files to: {log_dir}")
        return tensorboard_callback

```

Set up callbacks for training

```

In [21]: early_stopping = tf.keras.callbacks.EarlyStopping(monitor="val_loss", patience=3)
        reduce_lr = tf.keras.callbacks.ReduceLROnPlateau(monitor="val_loss", factor=0.2, patience=5)
        checkpoint_path = "fine_tune_checkpoints/"
        model_checkpoint = tf.keras.callbacks.ModelCheckpoint(
            checkpoint_path,
            save_weights_only=True,
            save_best_only=True,
            monitor="val_loss"
        )

```

Train the model with early stopping, learning rate reduction, and checkpointing

```

In [22]: epochs = 10
        history = feature_model.fit(train_data, epochs=epochs,
                                    steps_per_epoch=len(train_data),
                                    validation_data=test_data,
                                    validation_steps=len(test_data),
                                    callbacks=[early_stopping, model_checkpoint, reduce_lr,
                                              create_tensorboard_callback('Fruit_Vegetable_Diseases_Detection_Model/')]
        )

```

```

Saving TensorBoard log files to: Fruit_Vegetable_Diseases_Detection_Model/EfficientNetB0
10/20231115-180057
Epoch 1/10
WARNING:tensorflow:From C:\Users\PMLS\anaconda3\Lib\site-packages\keras\src\utils\tf_utils.py:492: The name tf.ragged.RaggedTensorValue is deprecated. Please use tf.compat.v1.ragged.RaggedTensorValue instead.

```

```

WARNING:tensorflow:From C:\Users\PMLS\anaconda3\Lib\site-packages\keras\src\engine\base_layer_utils.py:384: The name tf.executing_eagerly_outside_functions is deprecated. Please use tf.compat.v1.executing_eagerly_outside_functions instead.

```

```

1467/1467 [=====] - 1524s 1s/step - loss: 0.2617 - accuracy: 0.

```

```

9256 - val_loss: 0.0801 - val_accuracy: 0.9741 - lr: 1.0000e-04
Epoch 2/10
1467/1467 [=====] - 1520s 1s/step - loss: 0.0423 - accuracy: 0.
9873 - val_loss: 0.0436 - val_accuracy: 0.9864 - lr: 1.0000e-04
Epoch 3/10
1467/1467 [=====] - 1543s 1s/step - loss: 0.0186 - accuracy: 0.
9946 - val_loss: 0.0339 - val_accuracy: 0.9886 - lr: 1.0000e-04
Epoch 4/10
1467/1467 [=====] - 1498s 1s/step - loss: 0.0099 - accuracy: 0.
9974 - val_loss: 0.0392 - val_accuracy: 0.9882 - lr: 1.0000e-04
Epoch 5/10
1467/1467 [=====] - 1469s 1s/step - loss: 0.0065 - accuracy: 0.
9983 - val_loss: 0.0313 - val_accuracy: 0.9906 - lr: 1.0000e-04
Epoch 6/10
1467/1467 [=====] - 1462s 996ms/step - loss: 0.0061 - accuracy:
0.9984 - val_loss: 0.0356 - val_accuracy: 0.9886 - lr: 1.0000e-04
Epoch 7/10
1467/1467 [=====] - 1486s 1s/step - loss: 0.0051 - accuracy: 0.
9986 - val_loss: 0.0236 - val_accuracy: 0.9935 - lr: 1.0000e-04
Epoch 8/10
1467/1467 [=====] - 1479s 1s/step - loss: 0.0047 - accuracy: 0.
9986 - val_loss: 0.0273 - val_accuracy: 0.9927 - lr: 1.0000e-04
Epoch 9/10
1467/1467 [=====] - ETA: 0s - loss: 0.0020 - accuracy: 0.9995
Epoch 9: ReduceLROnPlateau reducing learning rate to 1.9999999494757503e-05.
1467/1467 [=====] - 1467s 1s/step - loss: 0.0020 - accuracy: 0.
9995 - val_loss: 0.0282 - val_accuracy: 0.9911 - lr: 1.0000e-04
Epoch 10/10
1467/1467 [=====] - 1470s 1s/step - loss: 6.9776e-04 - accurac
y: 0.9999 - val_loss: 0.0189 - val_accuracy: 0.9946 - lr: 2.0000e-05

```

Step 4: Model Evaluation

Load the best model checkpoint

```
In [23]: feature_model.load_weights(checkpoint_path)
```

```

WARNING:tensorflow:From C:\Users\PMLS\anaconda3\Lib\site-packages\keras\src\saving\legac
y\save.py:538: The name tf.train.NewCheckpointReader is deprecated. Please use tf.comp
t.v1.train.NewCheckpointReader instead.

```

```
Out[23]: <tensorflow.python.checkpoint.checkpoint.CheckpointLoadStatus at 0x26801769290>
```

Evaluate the model on the test data

```
In [24]: test_loss, test_accuracy = feature_model.evaluate(test_data)
```

```

367/367 [=====] - 270s 735ms/step - loss: 0.0189 - accuracy: 0.
9946

```

Print the evaluation results

```
In [25]: print(f"Test Loss: {test_loss:.2f}")
print(f"Test Accuracy: {test_accuracy * 100:.2f}%")
```

```

Test Loss: 0.02
Test Accuracy: 99.46%

```

Step 5: Data Visualization and Model Metrics

Define a function to plot training history

```
In [26]: def plot_history(history):
    loss = history.history['loss']
    val_loss = history.history['val_loss']
    epochs = history.epoch
    acc = history.history['accuracy']
    val_acc = history.history['val_accuracy']

    plt.figure(figsize=(10, 5))
    plt.subplot(1, 2, 1)
    plt.plot(epochs, loss, label='Training Loss')
    plt.plot(epochs, val_loss, label='Validation Loss')
    plt.title('Training and Validation Loss')
    plt.xlabel('Epoch')
    plt.legend()
    plt.grid(True)

    plt.subplot(1, 2, 2)
    plt.plot(epochs, acc, label='Training Accuracy')
    plt.plot(epochs, val_acc, label='Validation Accuracy')
    plt.title('Training and Validation Accuracy')
    plt.xlabel('Epoch')
    plt.legend()
    plt.grid(True)

    plt.tight_layout()
    plt.show()
```

Plot the training history

```
In [27]: plot_history(history)
```



Calculate additional metrics for model evaluation

```
In [28]: from sklearn.metrics import classification_report

def calculate_metrics(model, test_data):
    y_true = []
    y_pred = []
```

```

for images, labels in test_data:
    y_true.extend(np.argmax(labels, axis=1))
    y_pred.extend(np.argmax(model.predict(images), axis=1))

return y_true, y_pred

```

```
In [29]: y_true, y_pred = calculate_metrics(feature_model, test_data)
```

```

1/1 [=====] - 3s 3s/step
1/1 [=====] - 1s 781ms/step
1/1 [=====] - 1s 860ms/step
1/1 [=====] - 1s 742ms/step
1/1 [=====] - 1s 760ms/step
1/1 [=====] - 1s 704ms/step
1/1 [=====] - 1s 711ms/step
1/1 [=====] - 1s 783ms/step
1/1 [=====] - 1s 726ms/step
1/1 [=====] - 1s 824ms/step
1/1 [=====] - 1s 831ms/step
1/1 [=====] - 1s 825ms/step
1/1 [=====] - 1s 832ms/step
1/1 [=====] - 1s 871ms/step
1/1 [=====] - 1s 784ms/step
1/1 [=====] - 1s 776ms/step
1/1 [=====] - 1s 740ms/step
1/1 [=====] - 1s 731ms/step
1/1 [=====] - 1s 730ms/step
1/1 [=====] - 1s 719ms/step
1/1 [=====] - 1s 733ms/step
1/1 [=====] - 1s 799ms/step
1/1 [=====] - 1s 796ms/step
1/1 [=====] - 1s 733ms/step
1/1 [=====] - 1s 723ms/step
1/1 [=====] - 1s 788ms/step
1/1 [=====] - 1s 764ms/step
1/1 [=====] - 1s 793ms/step
1/1 [=====] - 1s 817ms/step
1/1 [=====] - 1s 786ms/step
1/1 [=====] - 1s 901ms/step
1/1 [=====] - 1s 838ms/step
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```

Print classification report

```
In [30]: print(classification_report(y_true, y_pred, target_names=class_names))
```

	precision	recall	f1-score	support
Apple_Healthy	1.00	1.00	1.00	488
Apple_Rotten	1.00	0.98	0.99	586
Banana_Healthy	1.00	1.00	1.00	399
Banana_Rotten	1.00	1.00	1.00	559
Bellpepper_Healthy	1.00	1.00	1.00	400
Bellpepper_Rotten	0.97	0.99	0.98	400
Carrot_Healthy	0.97	0.99	0.98	399
Carrot_Rotten	0.99	0.96	0.98	400
Cucumber_Healthy	0.99	1.00	1.00	400
Cucumber_Rotten	0.99	0.99	0.99	400
Grape_Healthy	1.00	1.00	1.00	400
Grape_Rotten	1.00	1.00	1.00	400

Guava__Healthy	1.00	1.00	1.00	400
Guava__Rotten	1.00	1.00	1.00	400
Jujube__Healthy	1.00	1.00	1.00	400
Jujube__Rotten	1.00	1.00	1.00	400
Mango__Healthy	0.99	0.99	0.99	400
Mango__Rotten	1.00	0.99	0.99	450
Orange__Healthy	1.00	1.00	1.00	415
Orange__Rotten	0.99	1.00	1.00	438
Pomegranate__Healthy	1.00	1.00	1.00	400
Pomegranate__Rotten	1.00	1.00	1.00	400
Potato__Healthy	0.99	0.99	0.99	400
Potato__Rotten	0.99	0.99	0.99	400
Strawberry__Healthy	1.00	1.00	1.00	400
Strawberry__Rotten	1.00	1.00	1.00	400
Tomato__Healthy	1.00	1.00	1.00	400
Tomato__Rotten	1.00	1.00	1.00	400
accuracy			0.99	11734
macro avg	0.99	0.99	0.99	11734
weighted avg	0.99	0.99	0.99	11734

Compute the confusion matrix

```
In [31]: confusion = confusion_matrix(y_true, y_pred)
```

Plot the confusion matrix

```
In [32]: from sklearn.metrics import ConfusionMatrixDisplay

def plot_confusion_matrix(confusion, class_names):
    num_classes = len(class_names)
    fig, ax = plt.subplots(figsize=(14, 14))

    # Convert the confusion matrix values to integers
    confusion = confusion.astype(int)

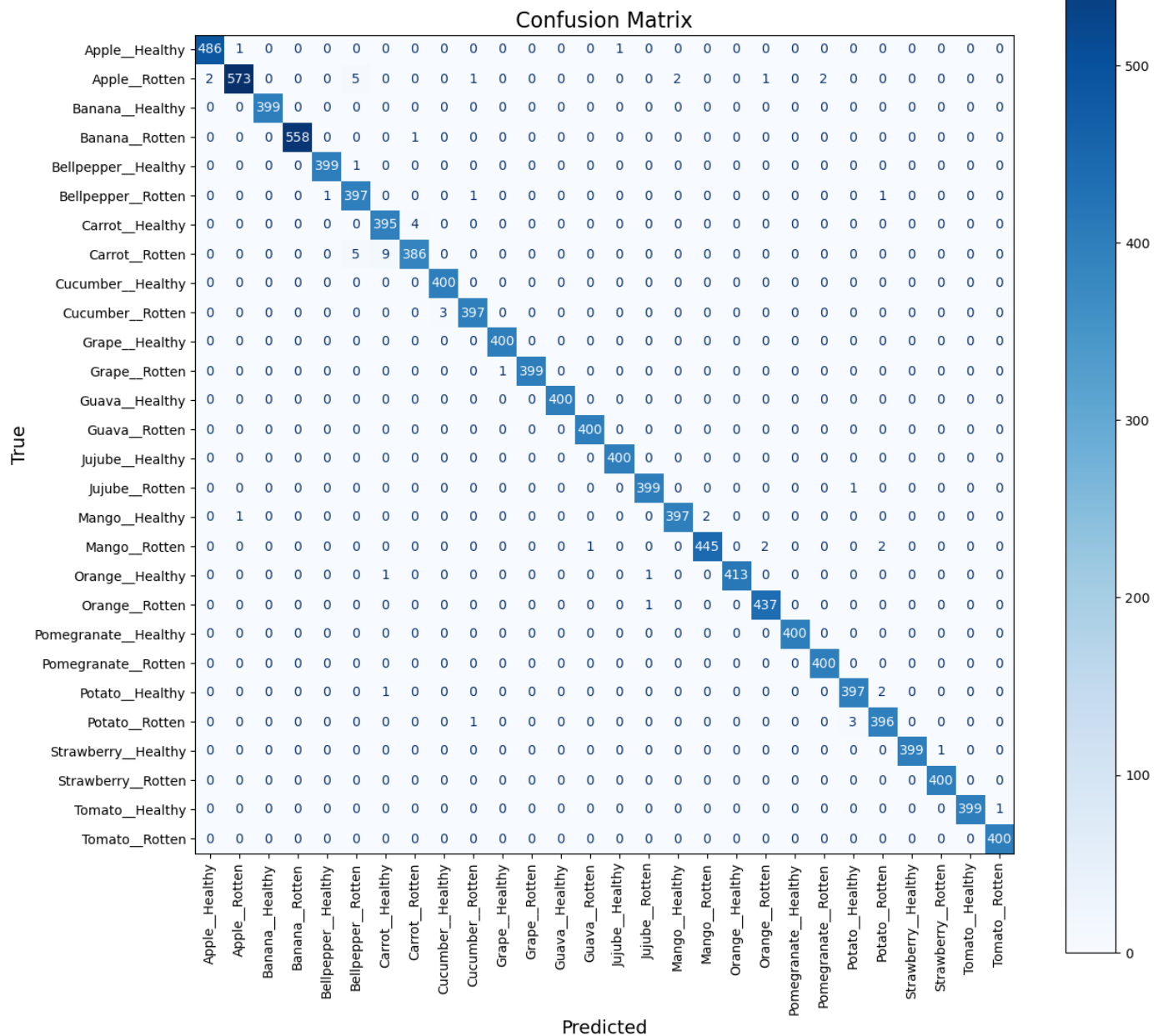
    disp = ConfusionMatrixDisplay(confusion, display_labels=class_names)
    disp = disp.plot(cmap=plt.get_cmap("Blues"), values_format="d", ax=ax)

    # Rotate y-axis class names to be straight at 90 degrees
    ax.set_yticklabels(class_names, rotation=0, fontsize=10)

    # Set the tick labels and fontsize for x-axis
    tick_marks = np.arange(num_classes)
    plt.xticks(tick_marks, class_names, rotation=90, fontsize=10)

    plt.title("Confusion Matrix", fontsize=16)
    plt.xlabel("Predicted", fontsize=14)
    plt.ylabel("True", fontsize=14)
    plt.show()

plot_confusion_matrix(confusion, class_names)
```



Step 6: Save Model

```
In [33]: tf.saved_model.save(feature_model, 'fruit_vegetable_disease_detection_model')
```

INFO:tensorflow:Assets written to: fruit_vegetable_disease_detection_model/assets

INFO:tensorflow:Assets written to: fruit_vegetable_disease_detection_model/assets

```
In [34]: h5_model_path = 'fruit_vegetable_disease_detection_model.h5'
feature_model.save(h5_model_path)
```

C:\Users\PMLS\anaconda3\Lib\site-packages\keras\src\engine\training.py:3103: UserWarning: You are saving your model as an HDF5 file via `model.save()`. This file format is considered legacy. We recommend using instead the native Keras format, e.g. `model.save('my_model.keras')`.

```
saving_api.save_model(
```

```
In [35]: tflite_model_path = 'fruit_vegetable_disease_detection_model.tflite'
converter = tf.lite.TFLiteConverter.from_keras_model(feature_model)
tflite_model = converter.convert()
```

```
with open(tflite_model_path, 'wb') as f:
    f.write(tflite_model)
```

INFO:tensorflow:Assets written to: C:\Users\PMLS\AppData\Local\Temp\tmpqgb4bq3y\assets

INFO:tensorflow:Assets written to: C:\Users\PMLS\AppData\Local\Temp\tmpqgb4bq3y\assets

Step 7: Image Prediction

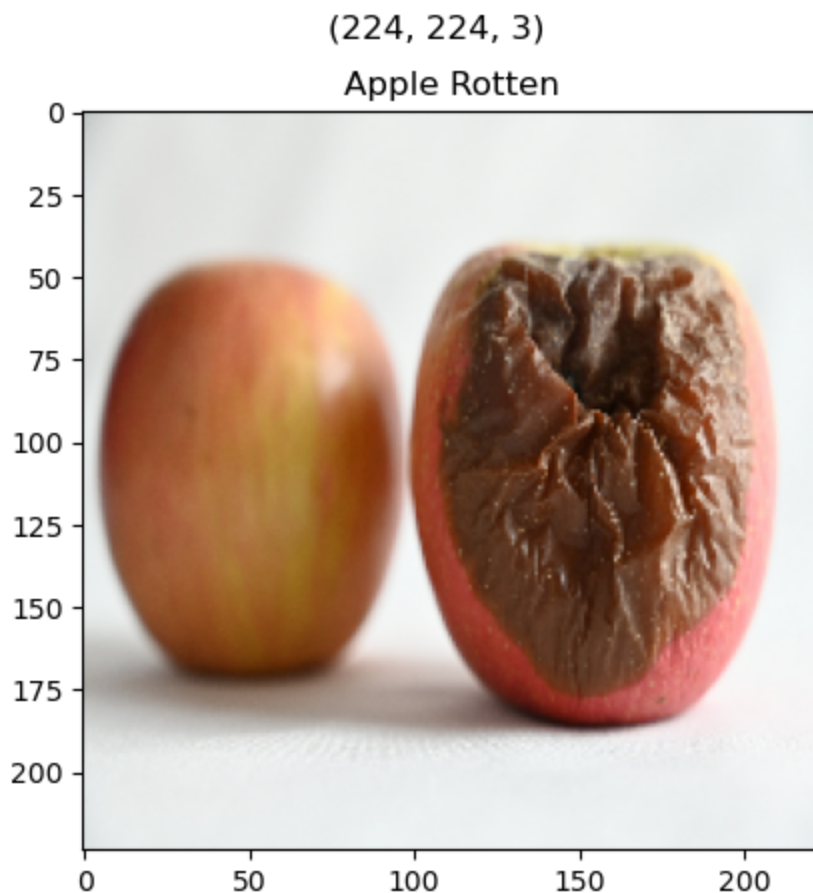
Define a function to load and preprocess an image

```
In [106... def load_prep(img_path):
    img = tf.io.read_file(img_path)
    img = tf.image.decode_image(img)
    img = tf.image.resize(img, size=(224, 224))
    return img
```

Load and preprocess an image, and make a prediction

```
In [107... image = load_prep('test/Apple Rotten 1.jpg')
plt.imshow(image / 255.)
plt.title('Apple Rotten')
plt.suptitle(image.shape)
```

Out[107]: Text(0.5, 0.98, '(224, 224, 3)')



```
In [108... pred = feature_model.predict(tf.expand_dims(image, axis=0))
predicted_class = class_names[pred.argmax()]
predicted_prob = pred.max()
```

1/1 [=====] - 1s 883ms/step

Print the predicted class and probability

```
In [109... print(f'Predicted Class: {predicted_class}')
print(f'Predicted Probability: {predicted_prob * 100:.2f}%')
```

Predicted Class: Apple__Rotten

Predicted Probability: 100.00%

Define a function to randomly select an image from the test data and make a prediction

```
In [110... def random_image_predict(model, test_dir=test_dir, class_names=class_names, rand_class=True):
    if rand_class:
        ran_cls = random.randint(0, len(class_names) - 1)
        cls = class_names[ran_cls]

        # Get a list of all files in the class directory
        class_dir = os.path.join(test_dir, cls)
        files = os.listdir(class_dir)

        # Choose a random file from the list
        random_file = random.choice(files)

        # Create the full path to the random file
        ran_path = os.path.join(class_dir, random_file)
    else:
        cls = class_names[cls_name]

        # Get a list of all files in the class directory
        class_dir = os.path.join(test_dir, cls)
        files = os.listdir(class_dir)

        # Choose a random file from the list
        random_file = random.choice(files)

        # Create the full path to the random file
        ran_path = os.path.join(class_dir, random_file)

    prep_img = load_prep(ran_path)

    pred = model.predict(tf.expand_dims(prepare_img, axis=0))
    pred_cls = class_names[pred[0].argmax()]
    pred_percent = pred[0][pred[0].argmax()] * 100
    plt.imshow(prepare_img / 255.)
    if pred_cls == cls:
        c = 'g'
    else:
        c = 'r'
    plt.title(f'Actual: {cls}\nPredicted: {pred_cls}\nProbability: {pred_percent:.2f}%',
              plt.axis(False))
```

Display 9 randomly predicted images from the test data

```
In [111... plt.figure(figsize=(15, 15))
for i in range(9):
    plt.subplot(3, 3, i + 1)
    random_image_predict(feature_model, test_dir)
```

```
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1/1 [=====] - 0s 231ms/step
```

Actual: Cucumber_Rotten
Predicted: Cucumber_Rotten
Probability: 100.00%



Actual: Mango_Healthy
Predicted: Mango_Healthy
Probability: 100.00%



Actual: Banana_Rotten
Predicted: Banana_Rotten
Probability: 100.00%



Actual: Cucumber_Healthy
Predicted: Cucumber_Healthy
Probability: 100.00%



Actual: Strawberry_Healthy
Predicted: Strawberry_Healthy
Probability: 99.97%



Actual: Bellpepper_Rotten
Predicted: Bellpepper_Rotten
Probability: 100.00%



Actual: Banana_Healthy
Predicted: Banana_Healthy
Probability: 100.00%



Actual: Cucumber_Healthy
Predicted: Cucumber_Healthy
Probability: 100.00%



Actual: Bellpepper_Rotten
Predicted: Bellpepper_Rotten
Probability: 100.00%



Define a directory containing images for prediction

```
In [112... data_dir = 'test'
plt.figure(figsize=(15, 10))
for i in range(9):
    plt.subplot(3, 3, i + 1)
    rn = random.choice(os.listdir(data_dir))
    image_path = os.path.join(data_dir, rn)
    img = load_prep(image_path)
    pred = feature_model.predict(tf.expand_dims(img, axis=0))
    pred_name = class_names[pred.argmax()]
    plt.imshow(img / 255.)
    plt.title(f'True: {rn}\nPredicted Class: {pred_name}')
    plt.axis(False)
```

```
1/1 [=====] - 0s 237ms/step
1/1 [=====] - 0s 249ms/step
1/1 [=====] - 0s 260ms/step
1/1 [=====] - 0s 298ms/step
```



```

1/1 [=====] - 0s 231ms/step
1/1 [=====] - 0s 216ms/step
1/1 [=====] - 0s 232ms/step
1/1 [=====] - 0s 272ms/step
1/1 [=====] - 0s 234ms/step

```

True: Grape Healthy 4.jpg
Predicted Class: Jujube_Healthy



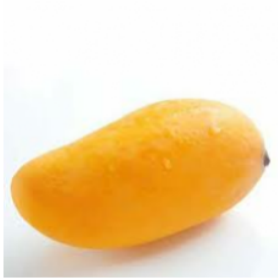
True: Potato Healthy 5.jpg
Predicted Class: Potato_Healthy



True: Potato Healthy 4.jpg
Predicted Class: Potato_Healthy



True: Mango Healthy 3.jpg
Predicted Class: Mango_Healthy



True: Tomato Healthy 6.jpg
Predicted Class: Tomato_Healthy



True: Tomato Healthy 6.jpg
Predicted Class: Tomato_Healthy



True: Apple Healthy 1.jpg
Predicted Class: Orange_Healthy



True: Potato Rotten 2.jpg
Predicted Class: Potato_Rotten



True: Cucumber Healthy 7.jpg
Predicted Class: Cucumber_Healthy



Define a function to predict an image from a given path

```

In [113... def predict_img(img_path, model=feature_model):
    img = load_prep(img_path)
    pred = model.predict(tf.expand_dims(img, axis=0))
    pred_name = class_names[pred.argmax()]
    plt.imshow(img / 255.)
    plt.title(f'Predicted Class: {pred_name}')
    plt.axis(False)

```

Step 9: Image Prediction for load Crop Diseases Detection model

```

In [141... loaded_model = tf.saved_model.load('fruit_vegetable_disease_detection_model')

```

Define a function to load and preprocess an image

```

In [147... def load_prep(img_path):
    img = tf.io.read_file(img_path)
    img = tf.image.decode_image(img)
    img = tf.image.resize(img, size=(224, 224))
    return img

```


Define the directory containing the images for prediction

```
In [148... test_directory = 'test'
```

Get a list of image file paths

```
In [149... image_paths = [os.path.join(test_directory, img) for img in os.listdir(test_directory)]
```

Make predictions on each image

```
In [150... predictions = []

for img_path in image_paths:
    img = load_prep(img_path)
    img = tf.expand_dims(img, axis=0)

    # Run inference using the loaded model
    prediction = loaded_model(img)
    predicted_class = class_names[np.argmax(prediction)]
    predictions.append((img_path, predicted_class))
```

Display the predictions

```
In [151... for img_path, predicted_class in predictions:
    print(f'Image: {os.path.basename(img_path)} - Predicted Class: {predicted_class}')
```

```
Image: Apple Healthy 1.jpg - Predicted Class: Orange_Healthy
Image: Apple Healthy 2.jpg - Predicted Class: Potato_Healthy
Image: Apple Healthy 3.jpg - Predicted Class: Apple_Healthy
Image: Apple Healthy 4.jpg - Predicted Class: Apple_Healthy
Image: Apple Healthy 5.jpg - Predicted Class: Strawberry_Healthy
Image: Apple Healthy 6.jpg - Predicted Class: Apple_Healthy
Image: Apple Rotten 1.jpg - Predicted Class: Apple_Rotten
Image: Apple Rotten 2.jpg - Predicted Class: Apple_Rotten
Image: Apple Rotten 3.jpg - Predicted Class: Apple_Rotten
Image: Apple Rotten 4.jpg - Predicted Class: Apple_Rotten
Image: Apple Rotten 5.jpg - Predicted Class: Tomato_Rotten
Image: Apple Rotten 6.jpg - Predicted Class: Apple_Rotten
Image: Banana Healthy 1.jpg - Predicted Class: Banana_Healthy
Image: Banana Healthy 2.jpg - Predicted Class: Banana_Healthy
Image: Banana Healthy 3.jpg - Predicted Class: Banana_Healthy
Image: Banana Healthy 4.jpg - Predicted Class: Banana_Healthy
Image: Banana Healthy 5.jpg - Predicted Class: Banana_Healthy
Image: Banana Healthy 6.jpg - Predicted Class: Banana_Healthy
Image: Banana Rotten 1.jpg - Predicted Class: Banana_Rotten
Image: Banana Rotten 2.jpg - Predicted Class: Banana_Rotten
Image: Banana Rotten 3.jpg - Predicted Class: Banana_Rotten
Image: Banana Rotten 4.jpg - Predicted Class: Banana_Rotten
Image: Banana Rotten 5.jpg - Predicted Class: Banana_Rotten
Image: Banana Rotten 6.jpg - Predicted Class: Banana_Rotten
Image: Bellpepper Healthy 1.jpg - Predicted Class: Bellpepper_Healthy
Image: Bellpepper Healthy 2.jpg - Predicted Class: Bellpepper_Healthy
Image: Bellpepper Healthy 3.jpg - Predicted Class: Bellpepper_Healthy
Image: Bellpepper Healthy 4.jpg - Predicted Class: Bellpepper_Healthy
Image: Bellpepper Healthy 5.jpg - Predicted Class: Bellpepper_Healthy
Image: Bellpepper Healthy 6.jpg - Predicted Class: Bellpepper_Rotten
Image: Bellpepper Healthy 7.jpg - Predicted Class: Bellpepper_Healthy
Image: Bellpepper Rotten 1.jpg - Predicted Class: Bellpepper_Rotten
Image: Bellpepper Rotten 2.jpg - Predicted Class: Bellpepper_Rotten
Image: Bellpepper Rotten 3.jpg - Predicted Class: Bellpepper_Rotten
Image: Bellpepper Rotten 4.jpg - Predicted Class: Bellpepper_Rotten
Image: Bellpepper Rotten 5.jpg - Predicted Class: Bellpepper_Rotten
```

Image: Bellpepper Rotten 6.jpg - Predicted Class: Bellpepper__Rotten
Image: Bellpepper Rotten 7.jpg - Predicted Class: Bellpepper__Rotten
Image: Carrot Healthy 2.jpg - Predicted Class: Carrot__Healthy
Image: Carrot Healthy 3.jpg - Predicted Class: Carrot__Healthy
Image: Carrot Healthy 4.jpg - Predicted Class: Carrot__Healthy
Image: Carrot Healthy 5.jpg - Predicted Class: Carrot__Healthy
Image: Carrot Healthy 6.jpg - Predicted Class: Carrot__Healthy
Image: Carrot Rotten 1.jpg - Predicted Class: Carrot__Rotten
Image: Carrot Rotten 2.jpg - Predicted Class: Carrot__Rotten
Image: Carrot Rotten 3.jpg - Predicted Class: Carrot__Rotten
Image: Carrot Rotten 4.jpg - Predicted Class: Carrot__Rotten
Image: Cucumber Healthy 4.jpg - Predicted Class: Cucumber__Healthy
Image: Cucumber Healthy 5.jpg - Predicted Class: Cucumber__Healthy
Image: Cucumber Healthy 6.jpg - Predicted Class: Cucumber__Healthy
Image: Cucumber Healthy 7.jpg - Predicted Class: Cucumber__Healthy
Image: Cucumber Healthy 1.jpg - Predicted Class: Cucumber__Healthy
Image: Cucumber Healthy 2.jpg - Predicted Class: Cucumber__Healthy
Image: Cucumber Healthy 3.jpg - Predicted Class: Cucumber__Healthy
Image: Cucumber Rotten 1.jpg - Predicted Class: Cucumber__Rotten
Image: Cucumber Rotten 2.jpg - Predicted Class: Cucumber__Rotten
Image: Cucumber Rotten 3.jpg - Predicted Class: Cucumber__Rotten
Image: Cucumber Rotten 4.jpg - Predicted Class: Cucumber__Rotten
Image: Cucumber Rotten 5.jpg - Predicted Class: Cucumber__Rotten
Image: Cucumber Rotten 6.jpg - Predicted Class: Cucumber__Rotten
Image: Grape Healthy 1.jpg - Predicted Class: Apple__Rotten
Image: Grape Healthy 2.jpg - Predicted Class: Grape__Healthy
Image: Grape Healthy 3.jpg - Predicted Class: Tomato__Healthy
Image: Grape Healthy 4.jpg - Predicted Class: Jujube__Healthy
Image: Grape Healthy 5.jpg - Predicted Class: Grape__Healthy
Image: Grape Healthy 6.jpg - Predicted Class: Jujube__Healthy
Image: Guava Healthy 1.jpg - Predicted Class: Guava__Healthy
Image: Guava Healthy 2.jpg - Predicted Class: Guava__Healthy
Image: Guava Healthy 3.jpg - Predicted Class: Cucumber__Rotten
Image: Guava Healthy 4.jpg - Predicted Class: Guava__Healthy
Image: Guava Rotten 1.jpg - Predicted Class: Apple__Rotten
Image: Guava Rotten 2.jpg - Predicted Class: Mango__Rotten
Image: Guava Rotten 3.jpg - Predicted Class: Guava__Rotten
Image: Guava Rotten 4.jpg - Predicted Class: Mango__Rotten
Image: Guava Rotten 5.jpg - Predicted Class: Apple__Rotten
Image: Jujube Healthy 1.jpg - Predicted Class: Tomato__Healthy
Image: Jujube Healthy 2.jpg - Predicted Class: Jujube__Healthy
Image: Jujube Healthy 3.jpg - Predicted Class: Jujube__Healthy
Image: Jujube Healthy 4.jpg - Predicted Class: Jujube__Healthy
Image: Jujube Healthy 5.jpg - Predicted Class: Jujube__Healthy
Image: Jujube Rotten 1.jpg - Predicted Class: Apple__Rotten
Image: Jujube Rotten 2.jpg - Predicted Class: Apple__Rotten
Image: Jujube Rotten 3.jpg - Predicted Class: Jujube__Rotten
Image: Jujube Rotten 4.jpg - Predicted Class: Jujube__Rotten
Image: Mango Healthy 1.jpg - Predicted Class: Mango__Healthy
Image: Mango Healthy 2.jpg - Predicted Class: Mango__Healthy
Image: Mango Healthy 3.jpg - Predicted Class: Mango__Healthy
Image: Mango Healthy 4.jpg - Predicted Class: Mango__Healthy
Image: Mango Rotten 1.jpg - Predicted Class: Mango__Rotten
Image: Mango Rotten 2.jpg - Predicted Class: Mango__Rotten
Image: Mango Rotten 3.jpg - Predicted Class: Mango__Rotten
Image: Mango Rotten 4.jpg - Predicted Class: Mango__Rotten
Image: Orange Healthy 1.png - Predicted Class: Orange__Healthy
Image: Orange Healthy 5.png - Predicted Class: Orange__Healthy
Image: Orange Rotten 1.jpg - Predicted Class: Orange__Rotten
Image: Orange Rotten 2.jpg - Predicted Class: Orange__Rotten
Image: Orange Rotten 3.jpg - Predicted Class: Orange__Rotten
Image: Orange Rotten 4.jpg - Predicted Class: Orange__Rotten
Image: Orange Rotten 5.jpg - Predicted Class: Orange__Rotten
Image: Orange Rotten 6.jpg - Predicted Class: Orange__Rotten
Image: Pomegranate Healthy 1.jpg - Predicted Class: Tomato__Rotten
Image: Pomegranate Healthy 2.jpg - Predicted Class: Pomegranate__Healthy

Image: Pomegranate Healthy 3.jpg - Predicted Class: Apple__Healthy
Image: Pomegranate Healthy 4.jpg - Predicted Class: Tomato__Rotten
Image: Pomegranate Rotten 1.jpg - Predicted Class: Pomegranate__Rotten
Image: Pomegranate Rotten 2.jpg - Predicted Class: Pomegranate__Rotten
Image: Pomegranate Rotten 3.jpg - Predicted Class: Pomegranate__Rotten
Image: Potato Healthy 1.jpg - Predicted Class: Potato__Healthy
Image: Potato Healthy 2.jpg - Predicted Class: Potato__Healthy
Image: Potato Healthy 3.jpg - Predicted Class: Potato__Healthy
Image: Potato Healthy 4.jpg - Predicted Class: Potato__Healthy
Image: Potato Healthy 5.jpg - Predicted Class: Potato__Healthy
Image: Potato Healthy 6.jpg - Predicted Class: Potato__Healthy
Image: Potato Rotten 1.jpg - Predicted Class: Potato__Rotten
Image: Potato Rotten 2.jpg - Predicted Class: Potato__Rotten
Image: Potato Rotten 3.jpg - Predicted Class: Potato__Rotten
Image: Strawberry Healthy 1.jpg - Predicted Class: Strawberry__Rotten
Image: Strawberry Healthy 2.jpg - Predicted Class: Strawberry__Rotten
Image: Strawberry Healthy 3.jpg - Predicted Class: Strawberry__Healthy
Image: Strawberry Healthy 4.jpg - Predicted Class: Strawberry__Rotten
Image: Strawberry Healthy 5.jpg - Predicted Class: Strawberry__Healthy
Image: Strawberry Rotten 1.jpg - Predicted Class: Strawberry__Rotten
Image: Strawberry Rotten 2.jpg - Predicted Class: Strawberry__Rotten
Image: Strawberry Rotten 3.jpg - Predicted Class: Strawberry__Rotten
Image: Strawberry Rotten 4.jpg - Predicted Class: Strawberry__Rotten
Image: Strawberry Rotten 5.jpg - Predicted Class: Potato__Rotten
Image: Strawberry Rotten 6.jpg - Predicted Class: Strawberry__Rotten
Image: Tomato Healthy 1.jpg - Predicted Class: Tomato__Healthy
Image: Tomato Healthy 2.jpg - Predicted Class: Tomato__Healthy
Image: Tomato Healthy 3.jpg - Predicted Class: Tomato__Healthy
Image: Tomato Healthy 4.jpg - Predicted Class: Tomato__Healthy
Image: Tomato Healthy 5.jpg - Predicted Class: Tomato__Healthy
Image: Tomato Healthy 6.jpg - Predicted Class: Tomato__Healthy
Image: Tomato Rotten 1.jpg - Predicted Class: Tomato__Rotten
Image: Tomato Rotten 2.jpg - Predicted Class: Tomato__Rotten
Image: Tomato Rotten 3.jpg - Predicted Class: Tomato__Rotten
Image: Tomato Rotten 4.jpg - Predicted Class: Tomato__Rotten
Image: Tomato Rotten 5.jpg - Predicted Class: Tomato__Rotten