```
import zipfile
from google.colab import drive
drive.mount('/content/drive/')
     Drive already mounted at /content/drive/: to attempt to forcibly remount, call drive.mount("/content/drive/", force remount=True).
path_ = "/content/drive/MyDrive/AI_Project/plant_notplant/Dataset"
import tensorflow as tf
from tensorflow import keras
 from tensorflow.keras import layers
from tensorflow.keras.preprocessing.image import ImageDataGenerator
import os
import shutil
import random
from sklearn.model_selection import train_test_split
# Set the path to the dataset folder
dataset_dir = path_
# Define the training and testing ratios
train ratio = 0.8
test_ratio = 0.2
# Define the paths to the training and testing directories
train_dir = 'Dataset/train'
test_dir = 'Dataset/test'
  Create the training and testing directories if they don't exist
if not os.path.exists(train_dir):
    os.makedirs(train_dir)
if not os.path.exists(test dir):
    os.makedirs(test_dir)
# Iterate through each class in the dataset directory
for class_name in os.listdir(dataset_dir):
    class_dir = os.path.join(dataset_dir, class_name)
     if os.path.isdir(class_dir):
         # Get the list of image filenames for this class
         images = os.listdir(class_dir)
         # Split the image filenames into training and testing sets
train_images, test_images = train_test_split(images, train_size=train_ratio, test_size=test_ratio)
         # Copy the training images to the appropriate directory
for image in train_images:
             src = os.path.join(class_dir, image)
dst = os.path.join(train_dir, class_name, image)
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         # Copy the testing images to the appropriate directory
         for image in test_images:
             src = os.path.join(class dir, image)
              dst = os.path.join(test_dir, class_name, image)
             if not os.path.exists(os.path.join(test dir, class name)):
                  os.makedirs(os.path.join(test_dir, class_name))
             shutil.copy(src, dst)
# Set the paths to the dataset folders
train_data_dir = 'Dataset/train
test_data_dir = 'Dataset/test'
# Set the image dimensions and batch size
img_width, img_height = 224, 224
batch_size = 32
# Define the data generators for training and testing data
train_datagen = ImageDataGenerator(
    rescale=1./255.
     shear_range=0.2,
     zoom_range=0.2,
     horizontal flip=True)
test_datagen = ImageDataGenerator(rescale=1./255)
train_generator = train_datagen.flow_from_directory(
    train data dir,
     target_size=(img_width, img_height),
    batch size=batch size,
    class_mode='binary')
test_generator = test_datagen.flow_from_directory(
    test_data_dir,
target_size=(img_width, img_height),
    batch_size=batch_size,
class_mode='binary')
     Found 2316 images belonging to 2 classes. Found 1498 images belonging to 2 classes.
# Build the CNN model
model = keras.Sequential([
     layers.Conv2D(16, 3, padding='same', activation='relu', input_shape=(img_width, img_height, 3)),
     lavers.MaxPooling2D().
     layers.Conv2D(32, 3, padding='same', activation='relu'),
    layers.MaxPooling2D(),
     layers.Conv2D(64, 3, padding='same', activation='relu'),
     layers.MaxPooling2D(),
     layers.Flatten(),
     layers.Dense(128, activation='relu'),
     layers.Dense(1, activation='sigmoid')
```

```
# Compile the model
model.compile(optimizer='adam',
              loss='binary crossentropy',
              metrics=['accuracy'])
# Train the model
history = model.fit(train_generator,
                    epochs=10,
                    validation_data=test_generator)
    Epoch 1/10
73/73 [===
Гэ
                                  ======= 1 - 116s 2s/step - loss: 0.3560 - accuracy: 0.8472 - val loss: 0.1891 - val accuracy: 0.9479
     Epoch 2/10
73/73 [===
                               ========] - 111s 2s/step - loss: 0.1790 - accuracy: 0.9404 - val_loss: 0.2221 - val_accuracy: 0.9099
     Epoch 3/10
     73/73 [===
Epoch 4/10
                                 =======] - 113s 2s/step - loss: 0.1663 - accuracy: 0.9439 - val_loss: 0.2133 - val_accuracy: 0.9212
     73/73 [====
Epoch 5/10
73/73 [====
                                :=======] - 113s 2s/step - loss: 0.1365 - accuracy: 0.9521 - val_loss: 0.2595 - val_accuracy: 0.9119
     Epoch 6/10
                                 73/73 [===
Epoch 7/10
     73/73 [====
Epoch 8/10
73/73 [====
Epoch 9/10
                                ======= 1 - 110s 2s/step - loss: 0.0996 - accuracy: 0.9655 - val loss: 0.0755 - val accuracy: 0.9733
                              :=======] - 110s 2s/step - loss: 0.0893 - accuracy: 0.9689 - val_loss: 0.0889 - val_accuracy: 0.9706
     73/73 [=
                           ========= ] - 109s 1s/step - loss: 0.0816 - accuracy: 0.9741 - val_loss: 0.0561 - val_accuracy: 0.9840
     Epoch 10/10
73/73 [====
                          :========= ] - 111s 2s/step - loss: 0.1031 - accuracy: 0.9650 - val loss: 0.0651 - val accuracy: 0.9753
# Evaluate the model on the test set
test_loss, test_acc = model.evaluate(test_generator)
print('Test accuracy:', test_acc)
                                      ====] - 35s 757ms/step - loss: 0.0651 - accuracy: 0.9753
     Test accuracy: 0.9753003716468811
# Save the trained model
model.save('plant_classifier_model.h5')
import matplotlib.pvplot as plt
# Plot the training and validation loss
plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'])
plt.title('Model Loss')
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# Plot the training and validation accuracy
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.title('Model Accuracy')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.legend(['Train', 'Val'])
plt.show()
```

```
Model Loss
                                                                                                          Train
             0.35
                                                                                                          Val
# import tensorflow as tf
# import numpy as np
from tensorflow.keras.preprocessing.image import load img, img to array
# from tensorflow.keras.models import load_model
# Set the path to the trained model file
# model_path = 'plant_classifier_model.h5
# Load the trained model
# model = load_model(model_path)
# Set the image dimensions
img_width, img_height = 224, 224
# Define the classify_image function
def classify_image(image_path):
      # Load the image
      # Load the image
img = load_img(image_path, target_size=(img_width, img_height))
# Convert the image to a numpy array
img_array = img_to_array(img)
# Reshape the array to match the input shape of the model
      # Weshape the array to match the input snape or the model
img_array = img_array.reshape((1, img_width, img_height, 3))
# Preprocess the image
img_array = img_array / 255.0
# Predict the class label
      prediction = model.predict(img_array)
# Map the class label to a human-readable string
      if prediction < 0.5:
return 'Not a plant'
            return 'Plant'
             ---1 1
img_path = "/content/jm.jpeg"
result = classify_image(img_path)
print(result)
       1/1 [======
Not a plant
                                                 ======] - 0s 21ms/step
                                                                Epoch
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```

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