```
import warnings
warnings.filterwarnings('ignore')
# Import necessary libraries
import tensorflow as tf
from tensorflow.keras.preprocessing.image import ImageDataGenerator
# Define dataset paths
train_dir = "C:\\Users\\snayi\\Desktop\\My Data Science\\GitHubMaal\\Fruit quality\\dataset\\train"
\begin{tabular}{ll} \hline $$\text{dir} = $$\text{C:}\begin{tabular}{ll} \hline $$\text{distable} & \text{Colored} & \text{C
# Data augmentation for training
train_datagen = ImageDataGenerator(
           rescale=1.0/255.0,
                                                                           # Normalize pixel values (0-1 range)
           rotation range=20,
                                                                            # Rotate images randomly by 20 degrees
           width_shift_range=0.2, # Shift width randomly
          height_shift_range=0.2, # Shift height randomly
           shear_range=0.2,
                                                                            # Apply shearing
           zoom range=0.2,
                                                                             # Randomly zoom in/out
          horizontal_flip=True  # Randomly flip images horizontally
)
# Only normalize for testing data
test_datagen = ImageDataGenerator(rescale=1.0/255.0)
# Load training dataset
train_generator = train_datagen.flow_from_directory(
           train dir.
           target_size=(150, 150), # Resize all images to 150x150
           batch size=32,
                                                                             # Use batches of 32 images
           class_mode='binary'
                                                                            # Binary classification: fresh (1) or rotten (0)
# Load testing dataset
test_generator = test_datagen.flow_from_directory(
          test dir,
           target_size=(150, 150),
          batch size=32.
          class_mode='binary'
 Found 10901 images belonging to 6 classes.
              Found 2698 images belonging to 6 classes.
from \ tensorflow. keras. preprocessing. image \ import \ Image Data Generator \ and \ an algorithms of the property of the 
# Create ImageDataGenerators for training and testing
train_datagen = ImageDataGenerator(rescale=1.0/255) # Normalize pixel values
test_datagen = ImageDataGenerator(rescale=1.0/255)
# Flow data from directories
train_generator = train_datagen.flow_from_directory(
          train dir,
           target_size=(150, 150), # Resize images to 150x150
          batch size=32,
           class_mode='categorical' # Use 'categorical' for multi-class classification
test_generator = test_datagen.flow_from_directory(
          test dir,
           target_size=(150, 150),
          batch_size=32,
           class_mode='categorical'
          Found 10901 images belonging to 6 classes.
              Found 2698 images belonging to 6 classes.
print(train generator.class indices) # Should display a dictionary of class labels
print(train_generator.samples)
                                                                                                     # Number of images found in the training directory
             {'freshapples': 0, 'freshbanana': 1, 'freshoranges': 2, 'rottenapples': 3, 'rottenbanana': 4, 'rottenoranges': 5}
              10901
from tensorflow.keras import Sequential
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense
model = Sequential([
           Conv2D(32, (3, 3), activation='relu', input_shape=(150, 150, 3)),
```

```
MaxPooling2D((2, 2)),
   Conv2D(64, (3, 3), activation='relu'),
    MaxPooling2D((2, 2)),
    Flatten(),
   Dense(128, activation='relu').
    Dense(6, activation='softmax') # 6 neurons for 6 classes
1)
model.compile(optimizer='adam',loss='categorical crossentropy',metrics=['accuracy'])
train_generator = train_datagen.flow_from_directory(train_dir,target_size=(150, 150),batch_size=32,class_mode='categorical')
test_generator = test_datagen.flow_from_directory(test_dir,target_size=(150, 150),batch_size=32,class_mode='categorical')
    Found 10901 images belonging to 6 classes.
     Found 2698 images belonging to 6 classes.
history = model.fit(train_generator,steps_per_epoch=train_generator.samples // train_generator.batch_size,validation_data=test_generator
    validation_steps=test_generator.samples // test_generator.batch_size,
    epochs=10
)
   Epoch 1/10
     340/340
                                — 122s 353ms/step - accuracy: 0.6214 - loss: 1.2311 - val_accuracy: 0.8992 - val_loss: 0.2841
     Epoch 2/10
     340/340
                                - 0s 192us/step - accuracy: 0.9688 - loss: 0.1634 - val_accuracy: 1.0000 - val_loss: 0.1303
     Enoch 3/10
     340/340
                                - 123s 358ms/step - accuracy: 0.9027 - loss: 0.2803 - val_accuracy: 0.8746 - val_loss: 0.3435
     Epoch 4/10
     340/340
                                - 0s 101us/step - accuracy: 0.8438 - loss: 0.3368 - val_accuracy: 0.8000 - val_loss: 0.6780
     Epoch 5/10
     340/340 -
                                - 123s 359ms/step - accuracy: 0.9392 - loss: 0.1697 - val_accuracy: 0.9174 - val_loss: 0.2265
     Epoch 6/10
     340/340 -
                                — 0s 132us/step - accuracy: 0.9375 - loss: 0.0997 - val_accuracy: 1.0000 - val_loss: 0.0703
     Epoch 7/10
     340/340 -
                                - 123s 357ms/step - accuracy: 0.9634 - loss: 0.1060 - val_accuracy: 0.9215 - val_loss: 0.2133
     Epoch 8/10
     340/340 -
                                - 0s 112us/step - accuracy: 0.9688 - loss: 0.0959 - val accuracy: 1.0000 - val loss: 0.1322
     Epoch 9/10
     340/340 -
                                — 128s 375ms/step - accuracy: 0.9678 - loss: 0.0943 - val accuracy: 0.9338 - val loss: 0.1877
     Enoch 10/10
     340/340
                                - 1s 168us/step - accuracy: 0.9375 - loss: 0.1005 - val_accuracy: 1.0000 - val_loss: 0.0139
import numpy as np
import cv2
from tkinter import Tk
from tkinter.filedialog import askopenfilename
# Save the entire model to a file
model.save("fruit_quality_model.h5")
# Load the saved model
model = tf.keras.models.load_model("fruit_quality_model.h5")
WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save_model(model)`. This file format is c
     WARNING:absl:Compiled the loaded model, but the compiled metrics have yet to be built. `model.compile_metrics` will be empty until j
image_path = "C:\\Users\\snayi\\Desktop\\My Data Science\\GitHubMaal\\Fruit quality\\dataset\\test\\rottenbanana\\rotated_by_15_Screen 
# Call the function
def preprocess_image(image_path):
    img = cv2.imread(image_path)
                                                     # Read image
    img = cv2.cvtColor(img, cv2.COLOR_BGR2RGB)
                                                     # Convert to RGB
    img = cv2.resize(img, (150, 150))
                                                     # Resize to 150x150
    img = img / 255.0
                                                     # Normalize pixel values
    img = np.expand_dims(img, axis=0)
                                                     # Add batch dimension
    return img
preprocessed_image = preprocess_image(image_path)
print(preprocessed_image.shape) # Check the shape of the preprocessed image
\rightarrow (1, 150, 150, 3)
model = tf.keras.models.load_model("fruit_quality_model.h5")
   WARNING:absl:Compiled the loaded model, but the compiled metrics have yet to be built. `model.compile_metrics` will be empty until )
```

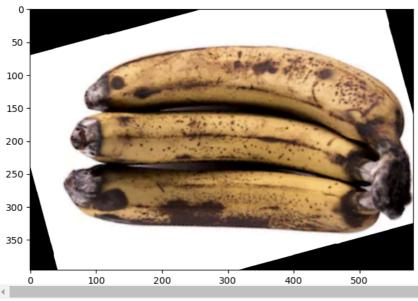
from tensorflow.keras.utils import load img, img to array

 $\verb|pil = load_img("C:\\\\)| Desktop\\ My Data Science\\ GitHubMaal\\ Fruit quality\\ dataset\\ test\\ rottenbanana\\ rotated_by_15_Screen Science\\ GitHubMaal\\ Fruit quality\\ GitHubMaal\\ Fruit quality\\$

from skimage.io import imshow

 $imshow ("C:\Users\svani) Desktop\My Data Science \GitHubMaal\Fruit quality\dataset\test\rottenbanana\rotated_by_15_Screen Shot 20 and 10 and$

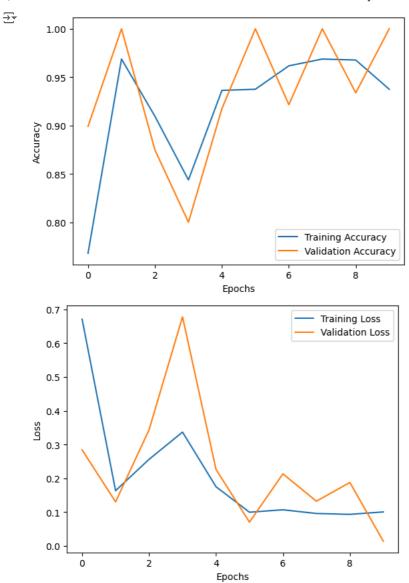




import matplotlib.pyplot as plt

```
# Plot training and validation accuracy
plt.plot(history.history['accuracy'], label='Training Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.xlabel('Epochs')
plt.ylabel('Accuracy')
plt.legend()
plt.show()

# Plot training and validation loss
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.legend()
plt.show()
```



Start coding or $\underline{\text{generate}}$ with AI.