

✓ Plant Disease Detection

importing Dependencies

```
import json
from zipfile import ZipFile
from PIL import Image

import numpy as np
import matplotlib.pyplot as plt
import matplotlib.image as mpimg

import tensorflow as tf
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras import layers, models

gpus = tf.config.experimental.list_physical_devices('GPU')
for gpu in gpus:
    tf.config.experimental.set_memory_growth(gpu, True)
```

Uploading Dataset

Data set link: <https://www.kaggle.com/datasets/abdallahalidev/plantvillage-dataset>

```
# Mounting Google Drive
from google.colab import drive
drive.mount('/content/drive')
```

↗ Mounted at /content/drive

```
# Unzipping the uploaded archive.zip file
zip_path = '/content/drive/My Drive/archive.zip'
with ZipFile(zip_path, 'r') as zip_ref:
    zip_ref.extractall('/content/plantvillage-dataset')
```

```
import os
print("Contents of '/content/plantvillage-dataset/New Plant Diseases Dataset(Augmented)':")
print(os.listdir('/content/plantvillage-dataset/New Plant Diseases Dataset(Augmented)'))
print("\nContents of '/content/plantvillage-dataset/new plant diseases dataset(augmented)':")
print(os.listdir('/content/plantvillage-dataset/new plant diseases dataset(augmented)'))
# Check contents of 'test'
print("\nContents of '/content/plantvillage-dataset/test':")
print(os.listdir('/content/plantvillage-dataset/test'))
```

↗ Contents of '/content/plantvillage-dataset/New Plant Diseases Dataset(Augmented)':
['New Plant Diseases Dataset(Augmented)']

Contents of '/content/plantvillage-dataset/new plant diseases dataset(augmented)':
['New Plant Diseases Dataset(Augmented)']

Contents of '/content/plantvillage-dataset/test':
['test']

```
import os

def find_image_folders(base_dir):
    image_extensions = ('.png', '.jpg', '.jpeg')
    folders_with_images = []

    for root, _, files in os.walk(base_dir):
        if any(file.lower().endswith(image_extensions) for file in files):
            folders_with_images.append(root)

    return folders_with_images

base_dir = '/content/plantvillage-dataset'
image_folders = find_image_folders(base_dir)

print("Folders containing image files:")
for folder in image_folders:
    print(folder)
```

[illegible]

```
df = '/content/plantvillage-dataset/New Plant Diseases Dataset(Augmented)/images'
```

Displaying Random 9 images

```

image_paths = [os.path.join(dp, f) for dp, dn, filenames in os.walk(df) for f in filenames if f.lower().endswith(('png', 'jpg', 'jpeg'))]

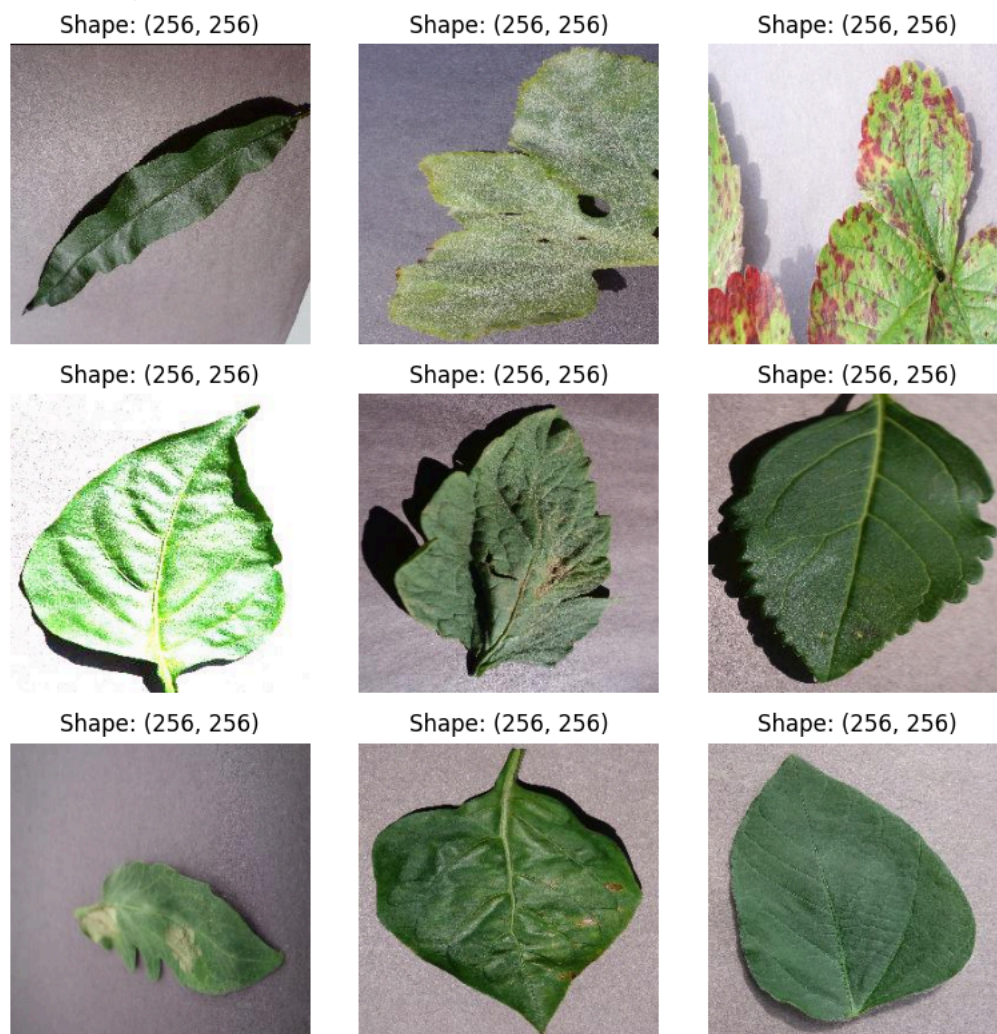
print(f"Number of images found: {len(image_paths)}")

if len(image_paths) < 9:
    print("Not enough images for sampling. Showing all available images instead.")
    selected_paths = image_paths
else:
    selected_paths = random.sample(image_paths, 9)

plt.figure(figsize=(8, 8))
for i, path in enumerate(selected_paths):
    img = Image.open(path)
    plt.subplot(3, 3, i + 1)
    plt.imshow(img)
    plt.title(f'Shape: {img.size}')
    plt.axis('off')
plt.tight_layout()
plt.show()

```

➡ Number of images found: 70295



```

img_size = 224
batch_size = 32

```

```

from tensorflow.keras.preprocessing.image import ImageDataGenerator

```

Normalization

```

data_gen = ImageDataGenerator(
    rescale=1./255,

```

```
validation_split=0.2
)

train_gen = data_gen.flow_from_directory(
    df,
    target_size=(img_size, img_size),
    batch_size=batch_size,
    subset='training',
    class_mode='categorical'
)

➡ Found 56251 images belonging to 38 classes.
```

```
validation_gen = data_gen.flow_from_directory(
    df,
    target_size=(img_size, img_size),
    batch_size=batch_size,
    subset='validation',
    class_mode='categorical'
)

➡ Found 14044 images belonging to 38 classes.
```

```
print(f"Training data batches: {len(train_gen)}")
print(f"Validation data batches: {len(validation_gen)}")

➡ Training data batches: 1758
   Validation data batches: 439
```

Model building

```
from tensorflow.keras import layers, models

model = models.Sequential([
    layers.Conv2D(16, (3, 3), activation='relu', input_shape=(img_size, img_size, 3)),
    layers.MaxPooling2D(2, 2),

    layers.Conv2D(32, (3, 3), activation='relu'),
    layers.MaxPooling2D(2, 2),

    layers.Flatten(),
    layers.Dense(128, activation='relu'),
    layers.Dense(train_gen.num_classes, activation='softmax') # Output layer with number of classes
])

➡ /usr/local/lib/python3.11/dist-packages/keras/src/layers/convolutional/base_conv.py:107: UserWarning: Do not pass an `input_shape` /
   super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```

```
model.summary()
```

➡ Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 222, 222, 16)	448
max_pooling2d (MaxPooling2D)	(None, 111, 111, 16)	0
conv2d_1 (Conv2D)	(None, 109, 109, 32)	4,640
max_pooling2d_1 (MaxPooling2D)	(None, 54, 54, 32)	0
flatten (Flatten)	(None, 93312)	0
dense (Dense)	(None, 128)	11,944,064
dense_1 (Dense)	(None, 38)	4,902

Total params: 11,954,054 (45.60 MB)

Training Model

```
model.compile(
```

```
optimizer='adam',
loss='categorical_crossentropy',
metrics=['accuracy']
)
```

```
history = model.fit(
    train_gen,
    epochs=5,
    validation_data=validation_gen
)
```

```
Epoch 1/5
/usr/local/lib/python3.11/dist-packages/keras/src/trainers/data_adapters/py_dataset_adapter.py:122: UserWarning: Your `PyDataset` class
    self._warn_if_super_not_called()
1758/1758 ————— 3048s 2s/step - accuracy: 0.5614 - loss: 1.7136 - val_accuracy: 0.8589 - val_loss: 0.4723
Epoch 2/5
1758/1758 ————— 3083s 2s/step - accuracy: 0.8978 - loss: 0.3194 - val_accuracy: 0.8559 - val_loss: 0.4815
Epoch 3/5
1758/1758 ————— 3087s 2s/step - accuracy: 0.9559 - loss: 0.1373 - val_accuracy: 0.8604 - val_loss: 0.5246
Epoch 4/5
1758/1758 ————— 3146s 2s/step - accuracy: 0.9747 - loss: 0.0810 - val_accuracy: 0.8671 - val_loss: 0.5566
Epoch 5/5
1758/1758 ————— 3138s 2s/step - accuracy: 0.9784 - loss: 0.0659 - val_accuracy: 0.8656 - val_loss: 0.5975
```

```
# Evaluate model performance
val_loss, val_accuracy = model.evaluate(validation_gen)
print(f"Validation Accuracy: {val_accuracy * 100:.2f}%")
```

```
439/439 ————— 226s 512ms/step - accuracy: 0.8660 - loss: 0.5997
Validation Accuracy: 86.56%
```

Visualization

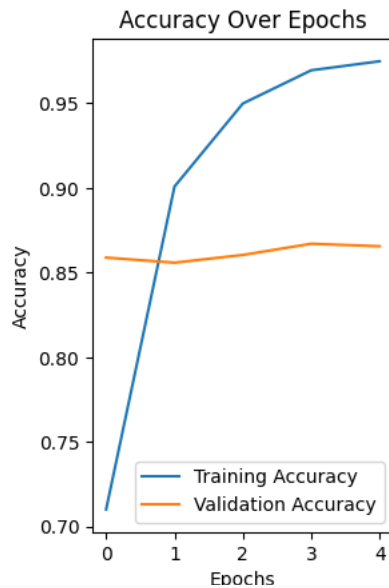
```
import matplotlib.pyplot as plt
```

```
plt.figure(figsize=(12, 5))
```

```
<Figure size 1200x500 with 0 Axes>
```

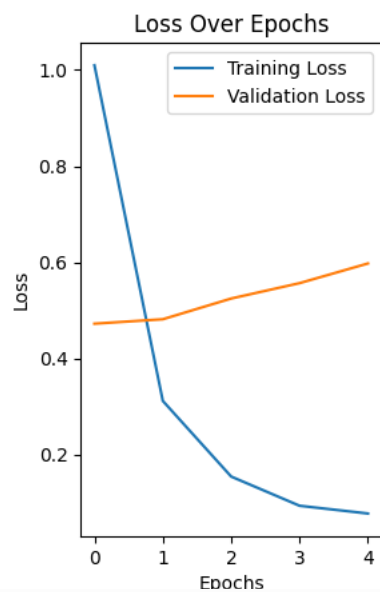
```
# Plotting accuracy
plt.subplot(1, 2, 1)
plt.plot(history.history['accuracy'], label='Training Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.title('Accuracy Over Epochs')
plt.xlabel('Epochs')
plt.ylabel('Accuracy')
plt.legend()
```

 <matplotlib.legend.Legend at 0x7bbab5109b90>



```
# Plotting loss
plt.subplot(1, 2, 2)
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.title('Loss Over Epochs')
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.legend()
```

 <matplotlib.legend.Legend at 0x7bbab54719d0>




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



```
predictions = model.predict(validation_gen, batch_size=batch_size)
predicted_labels = np.argmax(predictions, axis=1)
true_labels = validation_gen.classes
```

 439/439 — 232s 527ms/step

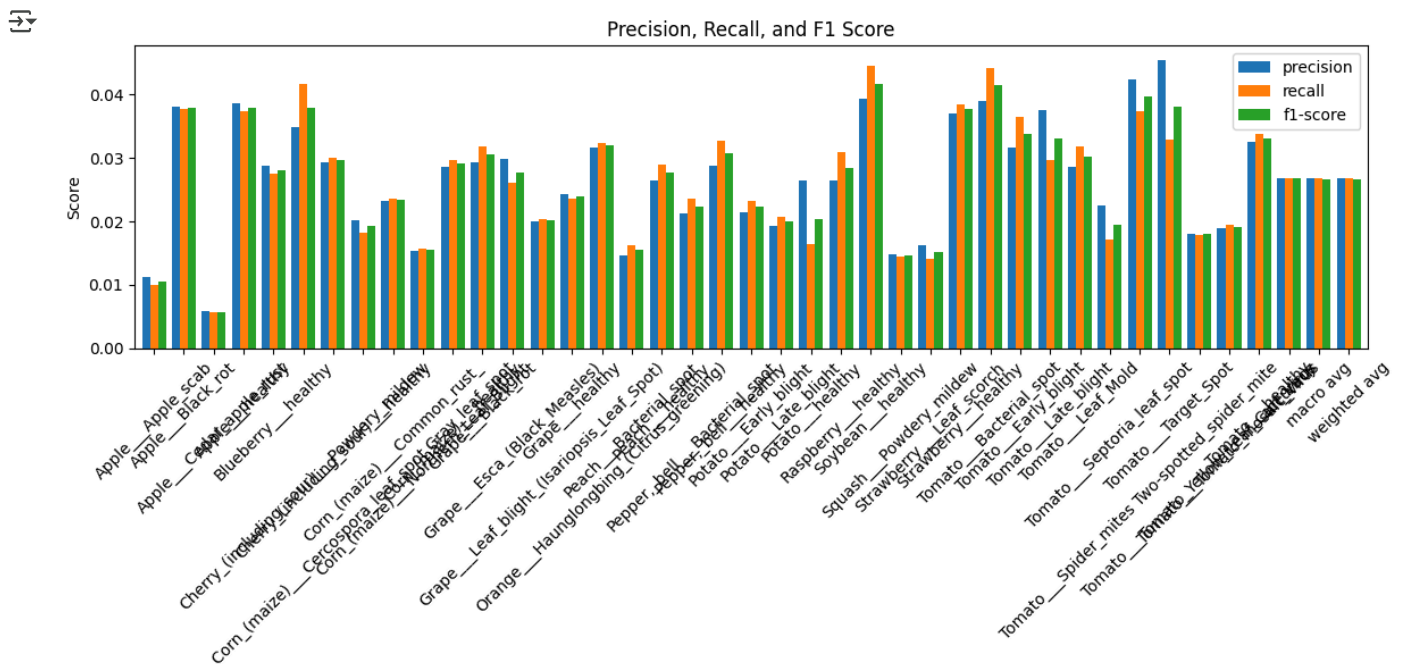
```
from sklearn.metrics import confusion_matrix, classification_report
report = classification_report(true_labels, predicted_labels, target_names=validation_gen.class_indices.keys())
print("Classification Report:\n", report)
```

 Classification Report:

precision recall f1-score support

Apple__Apple_scab	0.01	0.01	0.01	403
Apple__Black_rot	0.04	0.04	0.04	397
Apple__Cedar_apple_rust	0.01	0.01	0.01	352
Apple__healthy	0.04	0.04	0.04	401
Blueberry__healthy	0.03	0.03	0.03	363
Cherry_(including_sour)__Powdery_mildew	0.03	0.04	0.04	336
Cherry_(including_sour)__healthy	0.03	0.03	0.03	365
Corn_(maize)__Cercospora_leaf_spot_Gray_leaf_spot	0.02	0.02	0.02	328
Corn_(maize)__Common_rust	0.02	0.02	0.02	381
Corn_(maize)__Northern_Leaf_Blight	0.02	0.02	0.02	381
Corn_(maize)__healthy	0.03	0.03	0.03	371
Grape__Black_rot	0.03	0.03	0.03	377
Grape__Esca_(Black_Measles)	0.03	0.03	0.03	384
Grape__Leaf_blight_(Isariopsis_Leaf_Spot)	0.02	0.02	0.02	344
Grape__healthy	0.02	0.02	0.02	338
Orange__Haunglongbing_(Citrus_greening)	0.03	0.03	0.03	402
Peach__Bacterial_spot	0.01	0.02	0.02	367
Peach__healthy	0.03	0.03	0.03	345
Pepper,_bell__Bacterial_spot	0.02	0.02	0.02	382
Pepper,_bell__healthy	0.03	0.03	0.03	397
Potato__Early_blight	0.02	0.02	0.02	387
Potato__Late_blight	0.02	0.02	0.02	387
Potato__healthy	0.03	0.02	0.02	364
Raspberry__healthy	0.03	0.03	0.03	356
Soybean__healthy	0.04	0.04	0.04	404
Squash__Powdery_mildew	0.01	0.01	0.01	347
Strawberry__Leaf_scorch	0.02	0.01	0.02	354
Strawberry__healthy	0.04	0.04	0.04	364
Tomato__Bacterial_spot	0.04	0.04	0.04	340
Tomato__Early_blight	0.03	0.04	0.03	384
Tomato__Late_blight	0.04	0.03	0.03	370
Tomato__Leaf_Mold	0.03	0.03	0.03	376
Tomato__Septoria_leaf_spot	0.02	0.02	0.02	349
Tomato__Spider_mites_Two-spotted_spider_mite	0.04	0.04	0.04	348
Tomato__Target_Spot	0.05	0.03	0.04	365
Tomato__Tomato_Yellow_Leaf_Curl_Virus	0.02	0.02	0.02	392
Tomato__Tomato_mosaic_virus	0.02	0.02	0.02	358
Tomato__healthy	0.03	0.03	0.03	385
accuracy			0.03	14044
macro avg	0.03	0.03	0.03	14044
weighted avg	0.03	0.03	0.03	14044

```
df_report[['precision', 'recall', 'f1-score']].plot(kind='bar', figsize=(12, 6), width=0.8)
plt.title('Precision, Recall, and F1 Score')
plt.ylabel('Score')
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```



```
import pickle
```

```
with open('plant_disease_prediction_model.pkl', 'wb') as model_file:  
    pickle.dump(model, model_file)
```