## Seeding for reproducibility

```
# Set seeds for reproducibility
import random
random.seed(0)
import numpy as np
np.random.seed(0)
import tensorflow as tf
tf.random.set seed(0)
```

### Importing the dependencies

```
import os
import json
from zipfile import ZipFile
import PIL
from PIL import Image
import numpy as np
import matplotlib.pyplot as plt
import matplotlib.image as mpimg
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras import layers, models
print("TensorFlow Version:", tf. version )
print("NumPy Version:", np. version )
print("Pillow (PIL) Version:", PIL.__version__)
print("JSON Version:", json.__version__ if hasattr(json, "__version__") else "Built-in module, no version")
print("OS Module: Built-in, no version")
print("Random Module: Built-in, no version")
→ TensorFlow Version: 2.18.0
     NumPy Version: 2.0.2
     Pillow (PIL) Version: 11.1.0
     JSON Version: 2.0.9
     OS Module: Built-in, no version
     Random Module: Built-in, no version
```

### **Data Curation**

Upload the kaggle.json file

```
!pip install kaggle
```

```
Requirement already satisfied: kaggle in /usr/local/lib/python3.11/dist-packages (1.7.4.2)
Requirement already satisfied: bleach in /usr/local/lib/python3.11/dist-packages (from kaggle) (6.2.0)
Requirement already satisfied: certifi>=14.05.14 in /usr/local/lib/python3.11/dist-packages (from kaggle) (2025.1.31)
Requirement already satisfied: charset-normalizer in /usr/local/lib/python3.11/dist-packages (from kaggle) (3.4.1)
Requirement already satisfied: idna in /usr/local/lib/python3.11/dist-packages (from kaggle) (3.10)
```

```
Requirement already satisfied: protobuf in /usr/local/lib/python3.11/dist-packages (from kaggle) (5.29.4)
     Requirement already satisfied: python-dateutil>=2.5.3 in /usr/local/lib/python3.11/dist-packages (from kaggle) (2.8.2)
     Requirement already satisfied: python-slugify in /usr/local/lib/python3.11/dist-packages (from kaggle) (8.0.4)
     Requirement already satisfied: requests in /usr/local/lib/python3.11/dist-packages (from kaggle) (2.32.3)
     Requirement already satisfied: setuptools>=21.0.0 in /usr/local/lib/python3.11/dist-packages (from kaggle) (75.2.0)
     Requirement already satisfied: six>=1.10 in /usr/local/lib/python3.11/dist-packages (from kaggle) (1.17.0)
     Requirement already satisfied: text-unidecode in /usr/local/lib/python3.11/dist-packages (from kaggle) (1.3)
     Requirement already satisfied: tqdm in /usr/local/lib/python3.11/dist-packages (from kaggle) (4.67.1)
     Requirement already satisfied: urllib3>=1.15.1 in /usr/local/lib/python3.11/dist-packages (from kaggle) (2.3.0)
     Requirement already satisfied: webencodings in /usr/local/lib/python3.11/dist-packages (from kaggle) (0.5.1)
kaggle_credentails = json.load(open("//kaggle.json"))
# setup Kaggle API key as environment variables
os.environ['KAGGLE USERNAME'] = kaggle credentails["username"]
os.environ['KAGGLE_KEY'] = kaggle_credentails["key"]
!kaggle datasets download -d abdallahalidev/plantvillage-dataset
Dataset URL: <a href="https://www.kaggle.com/datasets/abdallahalidev/plantvillage-dataset">https://www.kaggle.com/datasets/abdallahalidev/plantvillage-dataset</a>
     License(s): CC-BY-NC-SA-4.0
!1s
     plantvillage-dataset.zip sample_data
# Unzip the downloaded dataset
with ZipFile("plantvillage-dataset.zip", 'r') as zip ref:
    zip ref.extractall()
print(os.listdir("plantvillage dataset"))
print(len(os.listdir("plantvillage dataset/segmented")))
print(os.listdir("plantvillage dataset/segmented")[:5])
print(len(os.listdir("plantvillage dataset/color")))
print(os.listdir("plantvillage dataset/color")[:5])
print(len(os.listdir("plantvillage dataset/grayscale")))
print(os.listdir("plantvillage dataset/grayscale")[:5])
     ['color', 'grayscale', 'segmented']
     ['Grape__Esca_(Black_Measles)', 'Corn_(maize)__Cercospora_leaf_spot Gray_leaf_spot', 'Grape__Black_rot', 'Tomato__Bacterial_spot', 'Corn_(maize)__healthy']
     ['Grape__Esca_(Black_Measles)', 'Corn_(maize)__Cercospora_leaf_spot Gray_leaf_spot', 'Grape__Black_rot', 'Tomato__Bacterial_spot', 'Corn_(maize)__healthy']
     ['Grape__Esca_(Black_Measles)', 'Corn_(maize)__Cercospora_leaf_spot Gray_leaf_spot', 'Grape__Black_rot', 'Tomato__Bacterial_spot', 'Corn_(maize)__healthy']
```

Number of Classes = 38

```
print(len(os.listdir("plantvillage dataset/color/Grape___healthy")))
print(os.listdir("plantvillage dataset/color/Grape___healthy")[:5])
```

**→** 423

['c197dfe9-44d6-4a7e-bb5a-75e2bf05380b\_\_\_Mt.N.V\_HL 6100.JPG', '9ceba66a-d7b0-4ed4-98c3-37d361517a90\_\_\_Mt.N.V\_HL 6147.JPG', 'c05f4201-5ab9-4bbd-b19e-c36515b7b3a9\_\_\_Mt.N.V\_HL 61

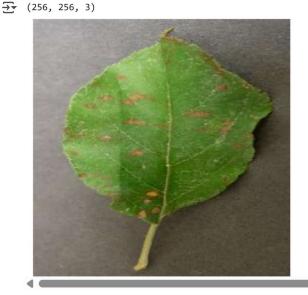
Data Preprocessing

```
# Dataset Path
base_dir = 'plantvillage dataset/color'

image_path = '/content/plantvillage dataset/color/Apple___Cedar_apple_rust/025b2b9a-0ec4-4132-96ac-7f2832d0db4a___FREC_C.Rust 3655.JPG'

# Read the image
img = mpimg.imread(image_path)

print(img.shape)
# Display the image
plt.imshow(img)
plt.axis('off') # Turn off axis numbers
plt.show()
```



image\_path = '\_/content/plantvillage dataset/color/Apple\_\_\_Cedar\_apple\_rust/025b2b9a-0ec4-4132-96ac-7f2832d0db4a\_\_\_FREC\_C.Rust 3655.JPG'
# Read the image
img = mpimg.imread(image\_path)

print(img)

```
→ [[[179 175 176]
      [181 177 178]
      [184 180 181]
      [115 112 105]
      [108 105 98]
      [101 98 91]]
     [[176 172 173]
      [177 173 174]
      [178 174 175]
      [113 110 103]
      [111 108 101]
      [109 106 99]]
     [[180 176 177]
      [180 176 177]
      [180 176 177]
      . . .
      [108 105 98]
      [111 108 101]
      [114 111 104]]
     [[137 128 119]
      [131 122 113]
      [125 116 107]
      . . .
      [ 74 65 48]
      [ 74 65 48]
      [ 73 64 47]]
     [[136 127 118]
      [132 123 114]
      [128 119 110]
      . . .
      [ 77 69 50]
      [ 75 67 48]
      [ 75 67 48]]
     [[133 124 115]
      [133 124 115]
      [132 123 114]
      . . .
      [ 81 73 54]
      [ 80 72 53]
      [ 79 71 52]]]
```

# Image Parameters
img\_size = 224

batch\_size = 32

## **Train Test Split**

```
# Image Data Generators
data_gen = ImageDataGenerator(
   rescale=1./255,
   validation_split=0.2 # Use 20% of data for validation
# Train Generator
train generator = data gen.flow from directory(
   base_dir,
   target_size=(img_size, img_size),
   batch_size=batch_size,
   subset='training',
   class mode='categorical'
    Found 43456 images belonging to 38 classes.
# Validation Generator
validation_generator = data_gen.flow_from_directory(
   base_dir,
   target_size=(img_size, img_size),
   batch_size=batch_size,
   subset='validation',
   class mode='categorical'
   Found 10849 images belonging to 38 classes.
Convolutional Neural Network
# Model Definition
model = models.Sequential()
model.add(layers.Conv2D(32, (3, 3), activation='relu', input shape=(img size, img size, 3)))
model.add(layers.MaxPooling2D(2, 2))
model.add(layers.Conv2D(64, (3, 3), activation='relu'))
model.add(layers.MaxPooling2D(2, 2))
model.add(layers.Flatten())
model.add(layers.Dense(256, activation='relu'))
model.add(layers.Dense(train generator.num classes, activation='softmax'))
super().__init__(activity_regularizer=activity_regularizer, **kwargs)
# model summary
model.summary()
```

```
→ Model: "sequential"
```

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 222, 222, 32)	896
max_pooling2d (MaxPooling2D)	(None, 111, 111, 32)	0
conv2d_1 (Conv2D)	(None, 109, 109, 64)	18,496
max_pooling2d_1 (MaxPooling2D)	(None, 54, 54, 64)	0
flatten (Flatten)	(None, 186624)	0
dense (Dense)	(None, 256)	47,776,000
dense_1 (Dense)	(None, 38)	9,766

```
Total params: 47,805,158 (182.36 MB)
```

### Model training

```
# Training the Model
history = model.fit(
    train generator,
    steps per epoch=train generator.samples // batch size, # Number of steps per epoch
    epochs=5, # Number of epochs
    validation_data=validation_generator,
    validation_steps=validation_generator.samples // batch_size # Validation steps

→ Epoch 1/5
     /usr/local/lib/python3.11/dist-packages/keras/src/trainers/data_adapters/py_dataset_adapter.py:121: UserWarning: Your `PyDataset` class should call `super().__init__(**kwargs)
       self._warn_if_super_not_called()
     1358/1358 -
                         ————— 107s 74ms/step - accuracy: 0.6062 - loss: 1.6166 - val accuracy: 0.8478 - val loss: 0.4724
     Epoch 2/5
                                 — 141s 104ms/step - accuracy: 0.9224 - loss: 0.2507 - val_accuracy: 0.8732 - val_loss: 0.4053
     1358/1358
     Epoch 3/5
     1358/1358
                                  - 94s 69ms/step - accuracy: 0.9671 - loss: 0.1021 - val accuracy: 0.8631 - val loss: 0.5516
     Epoch 4/5
                                  – 139s 67ms/step - accuracy: 0.9799 - loss: 0.0626 - val_accuracy: 0.8815 - val_loss: 0.4775
     1358/1358
     Epoch 5/5
```

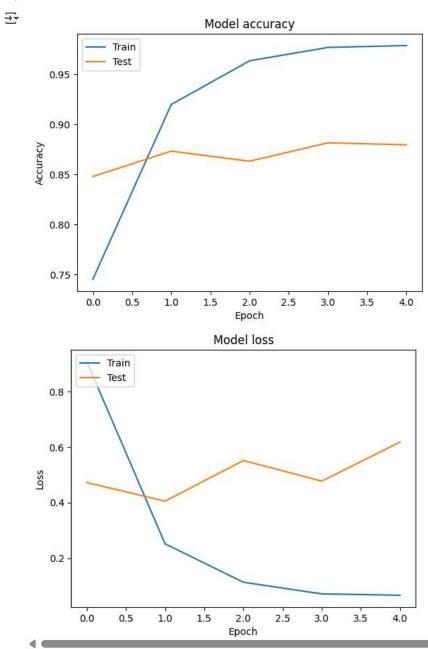
**— 142s** 67ms/step - accuracy: 0.9799 - loss: 0.0614 - val\_accuracy: 0.8795 - val\_loss: 0.6181

#### **Model Evaluation**

1358/1358

```
# Model Evaluation
print("Evaluating model...")
```

```
val_loss, val_accuracy = model.evaluate(validation_generator, steps=validation_generator.samples // batch_size)
print(f"Validation Accuracy: {val_accuracy * 100:.2f}%")
339/339 -
                          17s 50ms/step - accuracy: 0.8808 - loss: 0.6086
     Validation Accuracy: 87.96%
# Plot training & validation accuracy values
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.title('Model accuracy')
plt.ylabel('Accuracy')
plt.xlabel('Epoch')
plt.legend(['Train', 'Test'], loc='upper left')
plt.show()
# Plot training & validation loss values
plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'])
plt.title('Model loss')
plt.ylabel('Loss')
plt.xlabel('Epoch')
plt.legend(['Train', 'Test'], loc='upper left')
plt.show()
```



# **Building a Predictive System**

```
# Function to Load and Preprocess the Image using Pillow
def load_and_preprocess_image(image_path, target_size=(224, 224)):
    # Load the image
```

```
img = Image.open(image_path)
    # Resize the image
    img = img.resize(target size)
    # Convert the image to a numpy array
    img_array = np.array(img)
    # Add batch dimension
   img_array = np.expand_dims(img_array, axis=0)
    # Scale the image values to [0, 1]
    img_array = img_array.astype('float32') / 255.
    return img array
# Function to Predict the Class of an Image
def predict image class(model, image path, class indices):
    preprocessed img = load and preprocess image(image path)
    predictions = model.predict(preprocessed img)
    predicted class index = np.argmax(predictions, axis=1)[0]
    predicted_class_name = class_indices[predicted_class_index]
    return predicted_class_name
# Create a mapping from class indices to class names
class_indices = {v: k for k, v in train_generator.class_indices.items()}
class indices
→ {0: 'Apple Apple scab',
     1: 'Apple___Black_rot',
      2: 'Apple___Cedar_apple_rust',
     3: 'Apple___healthy',
     4: 'Blueberry__healthy',
      5: 'Cherry (including sour) Powdery mildew',
      6: 'Cherry_(including_sour)___healthy',
     7: 'Corn_(maize)___Cercospora_leaf_spot Gray_leaf_spot',
      8: 'Corn_(maize)___Common_rust_',
      9: 'Corn (maize) Northern Leaf Blight',
     10: 'Corn_(maize)___healthy',
      11: 'Grape___Black_rot',
      12: 'Grape___Esca_(Black_Measles)',
      13: 'Grape Leaf blight (Isariopsis Leaf Spot)',
      14: 'Grape___healthy',
      15: 'Orange___Haunglongbing_(Citrus_greening)',
      16: 'Peach___Bacterial_spot',
     17: 'Peach healthy',
      18: 'Pepper,_bell___Bacterial_spot',
      19: 'Pepper, bell healthy',
      20: 'Potato___Early_blight',
      21: 'Potato___Late_blight',
      22: 'Potato__healthy',
      23: 'Raspberry healthy',
      24: 'Soybean___healthy',
      25: 'Squash___Powdery_mildew'
      26: 'Strawberry___Leaf_scorch',
      27: 'Strawberry__healthy',
      28: 'Tomato___Bacterial_spot',
      29: 'Tomato___Early_blight',
      30: 'Tomato___Late_blight',
      31: 'Tomato___Leaf_Mold',
      32: 'Tomato Septoria leaf spot',
```

```
33: 'Tomato__Spider_mites Two-spotted_spider_mite',
34: 'Tomato__Target_Spot',
35: 'Tomato__Tomato_Yellow_Leaf_Curl_Virus',
36: 'Tomato__Tomato_mosaic_virus',
37: 'Tomato__healthy'}

# saving the class names as json file
json.dump(class_indices, open('class_indices.json', 'w'))

from tensorflow.python.saved_model.save import save
```