

```
In [2]: import tensorflow as tf
        from tensorflow.keras import models, layers
```

here we will load the data into tensor flow

```
In [4]: print(tf.config.list_physical_devices('GPU'))

[PhysicalDevice(name='/physical_device:GPU:0', device_type='GPU')]
```

```
In [7]: IMAGE_SIZE = 256
        BATCH_SIZE = 16
        CHANNELS = 3
        EPOCHS = 50
```

```
In [9]: dataset = tf.keras.preprocessing.image_dataset_from_directory(
        "PlantVillage/Plant",
        shuffle = True,
        image_size = (IMAGE_SIZE, IMAGE_SIZE),
        batch_size = BATCH_SIZE
    )
```

Found 13692 files belonging to 16 classes.

```
In [11]: def get_partition_and_data_tuning(ds, train_split= 0.8 , val_split=0.1, test_split = 0
        train_size = int(len(dataset)*train_split)
        val_size = int(len(dataset)*val_split)
        test_size = int(len(dataset)*test_split)
        if shuffle:
            ds = ds.shuffle(shuffle_size, seed = 12)
        train_data = ds.take(train_size)
        remain_data = ds.skip(train_size)
        val_data = remain_data.take(val_size)
        test_data = remain_data.skip(val_size)
        train_data = train_data.cache().shuffle(1000).prefetch(buffer_size=tf.data.AUTOTUNE)
        val_data = val_data.cache().shuffle(1000).prefetch(buffer_size=tf.data.AUTOTUNE)
        test_data = test_data.cache().shuffle(1000).prefetch(buffer_size=tf.data.AUTOTUNE)
        return train_data, val_data, test_data
```

```
In [13]: train, val, test = get_partition_and_data_tuning(dataset)
```

```
In [15]: resize_and_rescale = tf.keras.Sequential([

        layers.experimental.preprocessing.Resizing(IMAGE_SIZE, IMAGE_SIZE),
        layers.experimental.preprocessing.Rescaling(1.0/255),
    ])
```

```
In [17]: data_augmentation = tf.keras.Sequential([
        layers.experimental.preprocessing.RandomFlip("horizontal_and_vertical"),
        layers.experimental.preprocessing.RandomRotation(0.2),
    ])
```

```
In [19]: input=(IMAGE_SIZE,IMAGE_SIZE,CHANNELS)
        n_classes = 11
```

```
In [21]: model = tf.keras.Sequential([
        layers.Conv2D(32,(3,3),activation="relu",input_shape=input),
        layers.MaxPooling2D((2,2)),
        layers.Conv2D(64,(3,3),activation="relu"),
        layers.MaxPooling2D((2,2)),
        layers.Conv2D(64,(3,3),activation="relu"),
        layers.MaxPooling2D((2,2)),
        layers.Conv2D(64,(3,3),activation="relu"),
        layers.MaxPooling2D((2,2)),
        layers.Conv2D(64,(3,3),activation="relu"),
        layers.MaxPooling2D((2,2)),
        layers.Flatten(),
        layers.Dense(64,activation="relu"),
        layers.Dense(n_classes,activation="softmax")
    ])
```

```
In [23]: model.build(input_shape = input)
```

```
In [25]: model.summary()
```

Model: "sequential_2"

Layer (type)	Output Shape	Param #
=====		
conv2d (Conv2D)	(None, 254, 254, 32)	896
max_pooling2d (MaxPooling2D)	(None, 127, 127, 32)	0
conv2d_1 (Conv2D)	(None, 125, 125, 64)	18496
max_pooling2d_1 (MaxPooling2D)	(None, 62, 62, 64)	0
conv2d_2 (Conv2D)	(None, 60, 60, 64)	36928
max_pooling2d_2 (MaxPooling2D)	(None, 30, 30, 64)	0
conv2d_3 (Conv2D)	(None, 28, 28, 64)	36928
max_pooling2d_3 (MaxPooling2D)	(None, 14, 14, 64)	0
conv2d_4 (Conv2D)	(None, 12, 12, 64)	36928
max_pooling2d_4 (MaxPooling2D)	(None, 6, 6, 64)	0
conv2d_5 (Conv2D)	(None, 4, 4, 64)	36928
max_pooling2d_5 (MaxPooling2D)	(None, 2, 2, 64)	0
flatten (Flatten)	(None, 256)	0
dense (Dense)	(None, 64)	16448
dense_1 (Dense)	(None, 11)	715
=====		
Total params: 184,267		
Trainable params: 184,267		
Non-trainable params: 0		

```
In [ ]: with tf.device('/device:GPU:0'):  
        model.compile(  
            optimizer='adam',  
            loss=tf.keras.losses.SparseCategoricalCrossentropy(from_logits=False),  
            metrics=['accuracy']  
        )  
        history = model.fit(  
            train,  
            epochs = EPOCHS,  
            batch_size = BATCH_SIZE,
```

```
        verbose = 1,  
        validation_data = val  
    )
```

Epoch 1/50

271/684 [=====>.....] - ETA: 33s - loss: nan - accuracy: 0.0473

In [15]: `scores = model.evaluate(test)`

61/61 [=====] - 25s 53ms/step - loss: 0.1069 - accuracy: 0.9688

In [16]: `scores`

Out[16]: `[0.10689208656549454, 0.9688149690628052]`

In [17]: `history.params`

Out[17]: `{'verbose': 1, 'epochs': 50, 'steps': 482}`

In [18]: `history.history.keys()`

Out[18]: `dict_keys(['loss', 'accuracy', 'val_loss', 'val_accuracy'])`

In [19]: `len(history.history['accuracy'])`

Out[19]: `50`

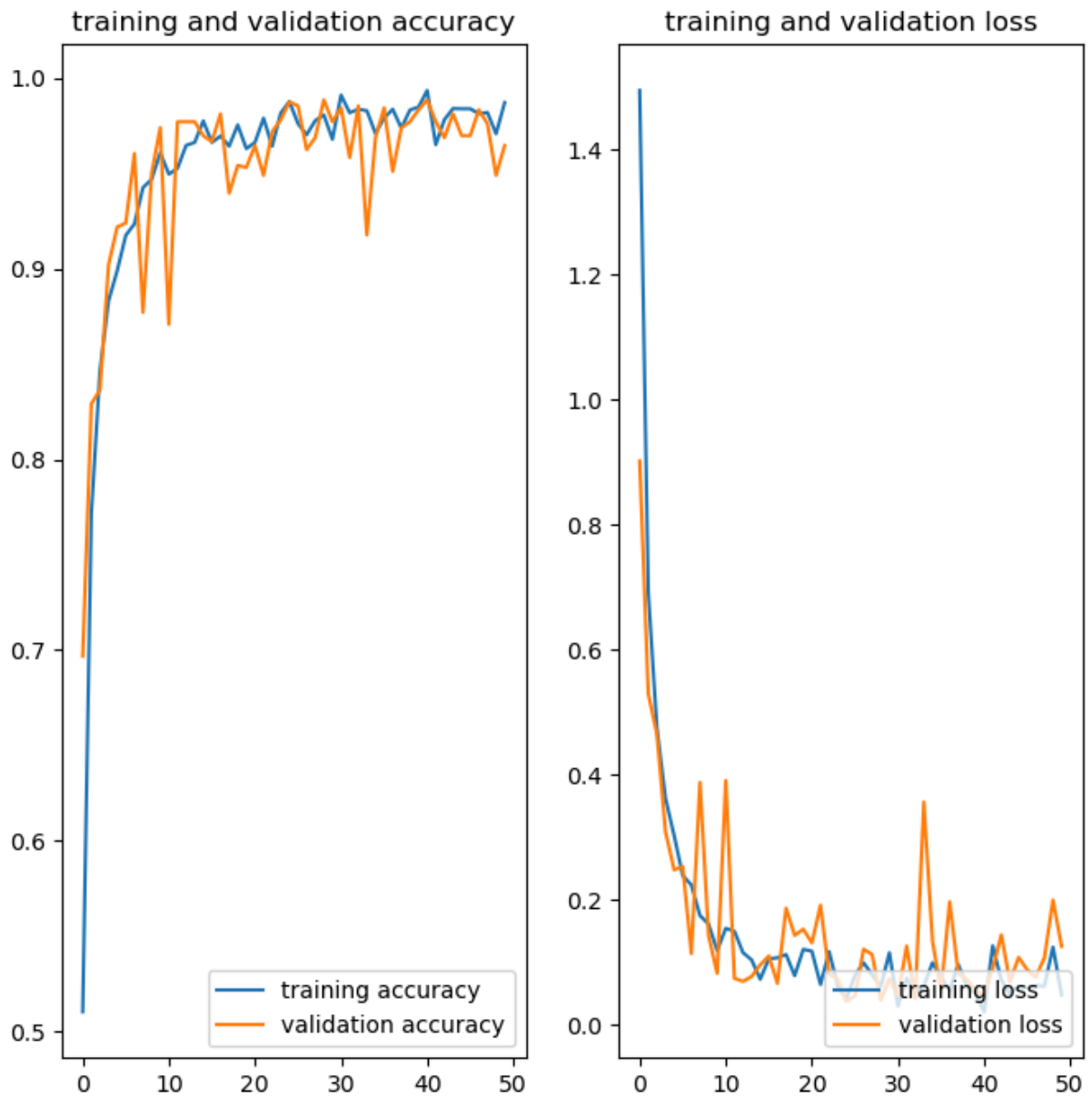
In [20]: `import matplotlib as mlt`

In [21]: `acc = history.history['accuracy']
val_acc = history.history['val_accuracy']
loss = history.history['loss']
val_loss = history.history['val_loss']`

In [22]: `plt.figure(figsize=(8,8))
plt.subplot(1,2,1)
plt.plot(range(EPOCHS),acc,label="training accuracy")
plt.plot(range(EPOCHS),val_acc,label="validation accuracy")
plt.legend(loc="lower right")
plt.title("training and validation accuracy")

plt.subplot(1,2,2)
plt.plot(range(EPOCHS),loss,label="training loss")
plt.plot(range(EPOCHS),val_loss,label="validation loss")
plt.legend(loc="lower right")
plt.title("training and validation loss")`

Out[22]: `Text(0.5, 1.0, 'training and validation loss')`



```
In [23]: import os
os.listdir("../leaf disease prediction/Models")
```

```
Out[23]: ['.ipynb_checkpoints',
'2',
'apple',
'apple_aware_model',
'cherry',
'cherry_aware_model',
'peach',
'peach_aware_model',
'pepper',
'pepper_aware_model',
'potato',
'potato_aware_model',
'quantize_aware_model',
'strawberry',
'strawberry_aware_model']
```

```
In [24]: # version = max([int(i) for i in os.listdir("../leaf disease prediction/Models")+
model.save(f"../leaf disease prediction/Models/quantize")
```

WARNING:absl:Found untraced functions such as _jit_compiled_convolution_op, _jit_compiled_convolution_op, _jit_compiled_convolution_op, _jit_compiled_convolution_op while saving (showing 5 of 6). These functions will not be directly callable after loading.

INFO:tensorflow:Assets written to: ../leaf disease prediction/Models/quantize/assets

INFO:tensorflow:Assets written to: ../leaf disease prediction/Models/quantize/assets

```
In [ ]:
```

```
In [25]: # _____predicting_____
```

```
In [26]: import numpy as np
```

```
In [27]: for images_batch, labels_batch in test.take(1):
    first_img = images_batch[0].numpy().astype('float32')
    first_label = labels_batch[0].numpy()
    print("first image to predict")
    plt.imshow(first_img)
    print("actual label:", dataset.class_names[first_label])
    batch_prediction = model.predict(images_batch)
    label1 = np.argmax(batch_prediction[0])
    print("predicted label:", dataset.class_names[label1])
```

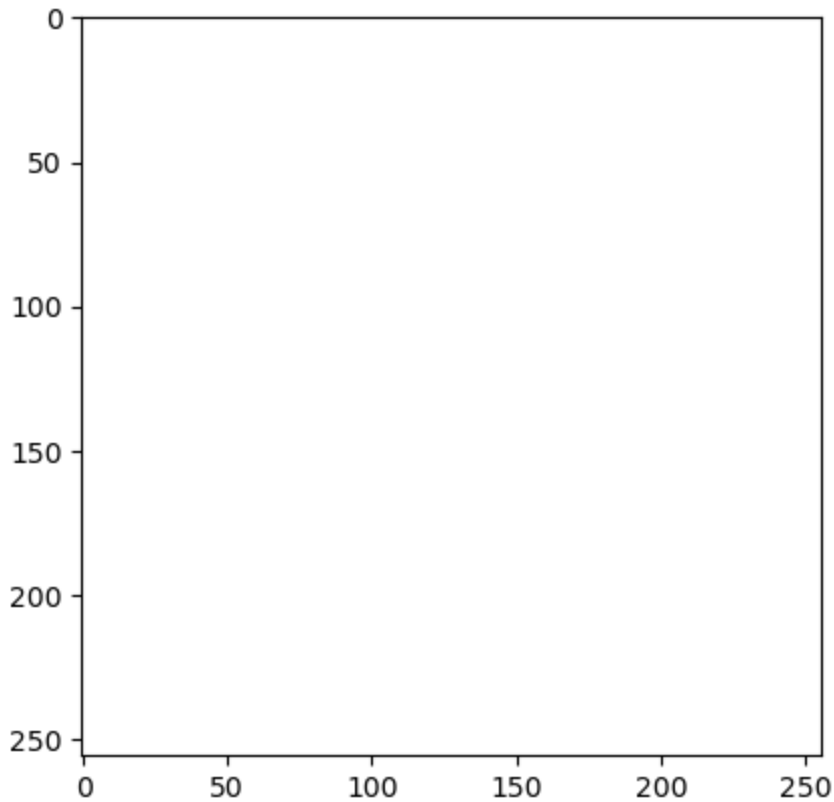
WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).

first image to predict

actual label: Apple__Cedar_apple_rust

1/1 [=====] - 0s 301ms/step

predicted label: Apple__Cedar_apple_rust



```
In [28]: def predict(model,img):  
  
    img_array = tf.keras.preprocessing.image.img_to_array(img)  
    img_array = tf.expand_dims(img_array,0)#create a batch  
    prediction = model.predict(img_array)  
    pre_class = dataset.class_names[np.argmax(prediction[0])]   
    confidence = round(100*(np.max(prediction[0])),2)  
    return pre_class,confidence
```

```
In [29]: #_____Loding and using the saved model_____
```

```
In [30]: from tensorflow.keras.models import load_model  
model = load_model("../leaf disease prediction/Models/quantize")
```

```
In [ ]:
```

```
In [31]: import tensorflow_model_optimization as tfmot
```

```
In [32]: new_model = tf.keras.models.load_model('Models/quantize')
```

```
In [33]: quantize_model = tfmot.quantization.keras.quantize_model
```

```
In [34]: quantize_aware_model = quantize_model(new_model)
```

```
In [35]: quantize_aware_model.compile(  
    optimizer='adam',  
    loss=tf.keras.losses.SparseCategoricalCrossentropy(from_logits=False),  
    metrics=['accuracy'])
```

```
In [36]: quantize_aware_model.summary()
```


Model: "sequential_2"

Layer (type)	Output Shape	Param #
quantize_layer (QuantizeLayer)	(None, 256, 256, 3)	3
quant_conv2d (QuantizeWrapperV2)	(None, 254, 254, 32)	963
quant_max_pooling2d (QuantizeWrapperV2)	(None, 127, 127, 32)	1
quant_conv2d_1 (QuantizeWrapperV2)	(None, 125, 125, 64)	18627
quant_max_pooling2d_1 (QuantizeWrapperV2)	(None, 62, 62, 64)	1
quant_conv2d_2 (QuantizeWrapperV2)	(None, 60, 60, 64)	37059
quant_max_pooling2d_2 (QuantizeWrapperV2)	(None, 30, 30, 64)	1
quant_conv2d_3 (QuantizeWrapperV2)	(None, 28, 28, 64)	37059
quant_max_pooling2d_3 (QuantizeWrapperV2)	(None, 14, 14, 64)	1
quant_conv2d_4 (QuantizeWrapperV2)	(None, 12, 12, 64)	37059
quant_max_pooling2d_4 (QuantizeWrapperV2)	(None, 6, 6, 64)	1
quant_conv2d_5 (QuantizeWrapperV2)	(None, 4, 4, 64)	37059
quant_max_pooling2d_5 (QuantizeWrapperV2)	(None, 2, 2, 64)	1
quant_flatten (QuantizeWrapperV2)	(None, 256)	1
quant_dense (QuantizeWrapperV2)	(None, 64)	16453
quant_dense_1 (QuantizeWrapperV2)	(None, 11)	720

Total params: 185,009

Trainable params: 184,267

Non-trainable params: 742

```
In [37]: with tf.device('/device:GPU:0'):
        quantize_aware_model.fit(
            train,
            epochs = 10,
            batch_size = BATCH_SIZE,
            verbose = 1,
            validation_data = val
        )
```

Epoch 1/10

482/482 [=====] - 108s 219ms/step - loss: 0.2550 - accuracy: 0.9224 - val_loss: 0.0998 - val_accuracy: 0.9719

Epoch 2/10

482/482 [=====] - 77s 159ms/step - loss: 0.0975 - accuracy: 0.9729 - val_loss: 0.2510 - val_accuracy: 0.9375

Epoch 3/10

482/482 [=====] - 69s 143ms/step - loss: 0.1286 - accuracy: 0.9725 - val_loss: 0.2399 - val_accuracy: 0.9521

Epoch 4/10

482/482 [=====] - 75s 154ms/step - loss: 0.0825 - accuracy: 0.9808 - val_loss: 0.0680 - val_accuracy: 0.9854

Epoch 5/10

482/482 [=====] - 68s 142ms/step - loss: 0.0583 - accuracy: 0.9834 - val_loss: 0.1412 - val_accuracy: 0.9688

Epoch 6/10

482/482 [=====] - 74s 154ms/step - loss: 0.0877 - accuracy: 0.9799 - val_loss: 0.1356 - val_accuracy: 0.9667

Epoch 7/10

482/482 [=====] - 79s 163ms/step - loss: 0.0495 - accuracy: 0.9856 - val_loss: 0.1833 - val_accuracy: 0.9490

Epoch 8/10

482/482 [=====] - 73s 151ms/step - loss: 0.0581 - accuracy: 0.9852 - val_loss: 0.1010 - val_accuracy: 0.9802

Epoch 9/10

482/482 [=====] - 74s 154ms/step - loss: 0.0562 - accuracy: 0.9845 - val_loss: 0.0740 - val_accuracy: 0.9802

Epoch 10/10

482/482 [=====] - 65s 134ms/step - loss: 0.0235 - accuracy: 0.9923 - val_loss: 0.0704 - val_accuracy: 0.9844

```
In [38]: ##### quantize_aware_model.evaluate(test)
        quantize_aware_model.save(f"../leaf disease prediction/Models/quantize_aware_model")
```

WARNING:absl:Found untraced functions such as conv2d_layer_call_fn, conv2d_layer_call_and_return_conditional_losses, _jit_compiled_convolution_op, conv2d_1_layer_call_fn, conv2d_1_layer_call_and_return_conditional_losses while saving (showing 5 of 24). These functions will not be directly callable after loading.

INFO:tensorflow:Assets written to: ../leaf disease prediction/Models/quantize_aware_model/assets

INFO:tensorflow:Assets written to: ../leaf disease prediction/Models/quantize_aware_model/assets

```
In [39]: convertor = tf.lite.TFLiteConverter.from_keras_model(quantize_aware_model)
convertor.optimizations = [tf.lite.Optimize.DEFAULT]
tf_model1 = convertor.convert()
```

WARNING:absl:Found untraced functions such as conv2d_layer_call_fn, conv2d_layer_call_and_return_conditional_losses, _jit_compiled_convolution_op, conv2d_1_layer_call_fn, conv2d_1_layer_call_and_return_conditional_losses while saving (showing 5 of 24). These functions will not be directly callable after loading.

INFO:tensorflow:Assets written to: C:\Users\heman\AppData\Local\Temp\tmpbpmpwed5\assets

INFO:tensorflow:Assets written to: C:\Users\heman\AppData\Local\Temp\tmpbpmpwed5\assets

C:\Users\heman\anaconda3\lib\site-packages\tensorflow\lite\python\convert.py:766: UserWarning: Statistics for quantized inputs were expected, but not specified; continuing anyway.

warnings.warn("Statistics for quantized inputs were expected, but not ")

```
In [40]: with open("quantize_tflite_model.tflite", "wb") as f:
f.write(tf_model1)
```

```
In [3]: nbconvert --to webpdf --allow-chromium-download modeltraining.ipynb
```

Cell In[3], line 1

```
nbconvert --to webpdf --allow-chromium-download modeltraining.ipynb
      ^
```

SyntaxError: invalid syntax

```
In [ ]:
```