Leaf Disease Classification - Keras CNN .. With 96% Accuracy

In [1]:

```
import warnings
warnings.filterwarnings("ignore")
import tensorflow as tf
import matplotlib.pyplot as plt
tf.compat.v1.set_random_seed(0)
from tensorflow import keras
import numpy as np
np.random.seed(0)
import itertools
from keras.preprocessing.image import image_dataset_from_directory
from tensorflow.keras.layers.experimental.preprocessing import Rescaling
from sklearn.metrics import precision_score, accuracy_score, recall_score, confusion_mat
```

Data Loading

Setting up Image Data Generators

Found 70295 files belonging to 38 classes.

1), but there must be at least one NUMA node, so returning NUMA node zero 2022-06-02 09:49:51.990947: I tensorflow/stream executor/cuda/cuda gpu exe cutor.cc:937] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-06-02 09:49:51.991830: I tensorflow/stream_executor/cuda/cuda_gpu_exe cutor.cc:937] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-06-02 09:49:51.994026: I tensorflow/core/platform/cpu_feature_guard.c c:142] This TensorFlow binary is optimized with oneAPI Deep Neural Network Library (oneDNN) to use the following CPU instructions in performance-crit ical operations: AVX2 AVX512F FMA To enable them in other operations, rebuild TensorFlow with the appropriat e compiler flags. 2022-06-02 09:49:51.994343: I tensorflow/stream_executor/cuda/cuda_gpu_exe cutor.cc:937] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-06-02 09:49:51.995141: I tensorflow/stream_executor/cuda/cuda_gpu_exe cutor.cc:937] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-06-02 09:49:51.995803: I tensorflow/stream_executor/cuda/cuda_gpu_exe cutor.cc:937] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-06-02 09:49:54.246926: I tensorflow/stream_executor/cuda/cuda_gpu_exe cutor.cc:937] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-06-02 09:49:54.247814: I tensorflow/stream executor/cuda/cuda gpu exe cutor.cc:937] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-06-02 09:49:54.248494: I tensorflow/stream_executor/cuda/cuda_gpu_exe cutor.cc:937] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero 2022-06-02 09:49:54.249073: I tensorflow/core/common_runtime/gpu/gpu_devic e.cc:1510] Created device /job:localhost/replica:0/task:0/device:GPU:0 wit h 15403 MB memory: -> device: 0, name: Tesla P100-PCIE-16GB, pci bus id: 0000:00:04.0, compute capability: 6.0

2022-06-02 09:49:51.830329: I tensorflow/stream_executor/cuda/cuda_gpu_exe cutor.cc:937] successful NUMA node read from SysFS had negative value (-

Found 17572 files belonging to 38 classes.

Data Engineering

Since the data is already augmented, there is no requirement of data engineering. Feature scaling is automatically done by image generators

Modelling

In [3]:

```
model = keras.Sequential()
model.add(keras.layers.Conv2D(32,(3,3),activation="relu",padding="same",input_shape=(256
model.add(keras.layers.Conv2D(32,(3,3),activation="relu",padding="same"))
model.add(keras.layers.MaxPooling2D(3,3))
model.add(keras.layers.Conv2D(64,(3,3),activation="relu",padding="same"))
model.add(keras.layers.Conv2D(64,(3,3),activation="relu",padding="same"))
model.add(keras.layers.MaxPooling2D(3,3))
model.add(keras.layers.Conv2D(128,(3,3),activation="relu",padding="same"))
model.add(keras.layers.Conv2D(128,(3,3),activation="relu",padding="same"))
model.add(keras.layers.MaxPooling2D(3,3))
model.add(keras.layers.Conv2D(256,(3,3),activation="relu",padding="same"))
model.add(keras.layers.Conv2D(256,(3,3),activation="relu",padding="same"))
model.add(keras.layers.Conv2D(512,(5,5),activation="relu",padding="same"))
model.add(keras.layers.Conv2D(512,(5,5),activation="relu",padding="same"))
model.add(keras.layers.Flatten())
model.add(keras.layers.Dense(1568,activation="relu"))
model.add(keras.layers.Dropout(0.5))
model.add(keras.layers.Dense(38,activation="softmax"))
opt = keras.optimizers.Adam(learning rate=0.0001)
model.compile(optimizer=opt,loss="sparse_categorical_crossentropy",metrics=['accuracy'])
model.summary()
```

Model: "sequential"

Layer (type)	Output	Shape	Param #
======================================	(None,	256, 256, 32)	896
conv2d_1 (Conv2D)	(None,	256, 256, 32)	9248
max_pooling2d (MaxPooling2D)	(None,	85, 85, 32)	0
conv2d_2 (Conv2D)	(None,	85, 85, 64)	18496
conv2d_3 (Conv2D)	(None,	85, 85, 64)	36928
max_pooling2d_1 (MaxPooling2	(None,	28, 28, 64)	0
conv2d_4 (Conv2D)	(None,	28, 28, 128)	73856
conv2d_5 (Conv2D)	(None,	28, 28, 128)	147584
max_pooling2d_2 (MaxPooling2	(None,	9, 9, 128)	0
conv2d_6 (Conv2D)	(None,	9, 9, 256)	295168
conv2d_7 (Conv2D)	(None,	9, 9, 256)	590080
conv2d_8 (Conv2D)	(None,	9, 9, 512)	3277312
conv2d_9 (Conv2D)	(None,	9, 9, 512)	6554112
flatten (Flatten)	(None,	41472)	0
dense (Dense)	(None,	1568)	65029664
dropout (Dropout)	(None,	1568)	0
	(None,	38)	59622

Total params: 76,092,966 Trainable params: 76,092,966

Non-trainable params: 0

```
In [4]:
ep = 10
history = model.fit_generator(train_gen,
       validation_data=test_gen,
       epochs = ep)
Epoch 1/10
2022-06-02 09:49:57.725110: I tensorflow/compiler/mlir/mlir graph optimiza
tion_pass.cc:185] None of the MLIR Optimization Passes are enabled (regist
ered 2)
2022-06-02 09:50:00.285178: I tensorflow/stream executor/cuda/cuda dnn.cc:
369] Loaded cuDNN version 8005
2 - accuracy: 0.5339 - val_loss: 0.5197 - val_accuracy: 0.8419
Epoch 2/10
2197/2197 [============== ] - 195s 89ms/step - loss: 0.4302
- accuracy: 0.8633 - val_loss: 0.2852 - val_accuracy: 0.9066
Epoch 3/10
2197/2197 [=============== ] - 194s 88ms/step - loss: 0.2553
- accuracy: 0.9166 - val_loss: 0.2539 - val_accuracy: 0.9214
Epoch 4/10
- accuracy: 0.9418 - val_loss: 0.2177 - val_accuracy: 0.9314
Epoch 5/10
2197/2197 [=============== ] - 193s 88ms/step - loss: 0.1299
- accuracy: 0.9573 - val_loss: 0.1694 - val_accuracy: 0.9451
Epoch 6/10
- accuracy: 0.9661 - val_loss: 0.1315 - val_accuracy: 0.9583
Epoch 7/10
- accuracy: 0.9719 - val_loss: 0.1392 - val_accuracy: 0.9585
Epoch 8/10
- accuracy: 0.9769 - val_loss: 0.1477 - val_accuracy: 0.9547
Epoch 9/10
2197/2197 [============= ] - 194s 88ms/step - loss: 0.0623
```

- accuracy: 0.9796 - val loss: 0.1131 - val accuracy: 0.9668

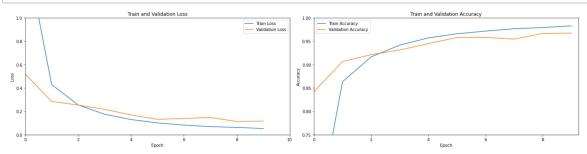
- accuracy: 0.9829 - val_loss: 0.1166 - val_accuracy: 0.9677

Metrics

Epoch 10/10

In [5]:

```
plt.figure(figsize = (20,5))
plt.subplot(1,2,1)
plt.title("Train and Validation Loss")
plt.xlabel("Epoch")
plt.ylabel("Loss")
plt.plot(history.history['loss'],label="Train Loss")
plt.plot(history.history['val_loss'], label="Validation Loss")
plt.xlim(0, 10)
plt.ylim(0.0,1.0)
plt.legend()
plt.subplot(1,2,2)
plt.title("Train and Validation Accuracy")
plt.xlabel("Epoch")
plt.ylabel("Accuracy")
plt.plot(history.history['accuracy'], label="Train Accuracy")
plt.plot(history.history['val accuracy'], label="Validation Accuracy")
plt.xlim(0, 9.25)
plt.ylim(0.75,1.0)
plt.legend()
plt.tight_layout()
```



In [6]:

```
labels = []
predictions = []
for x,y in test_gen:
    labels.append(list(y.numpy()))
    predictions.append(tf.argmax(model.predict(x),1).numpy())
```

In [7]:

```
predictions = list(itertools.chain.from_iterable(predictions))
labels = list(itertools.chain.from_iterable(labels))
```

In [8]:

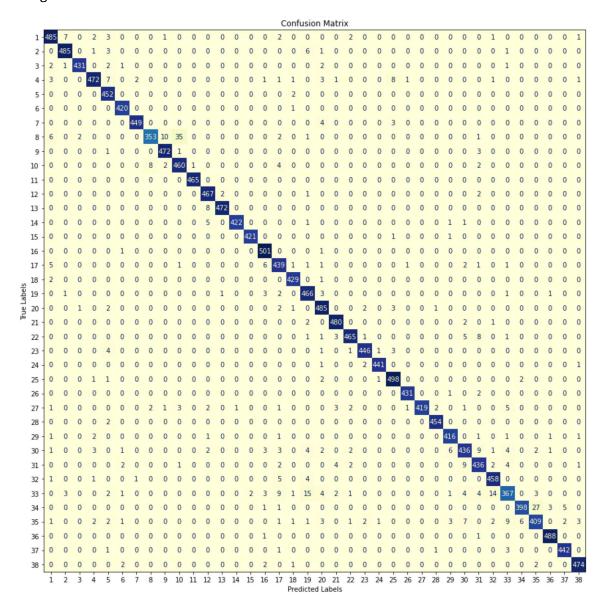
```
print("Train Accuracy : {:.2f} %".format(history.history['accuracy'][-1]*100))
print("Test Accuracy : {:.2f} %".format(accuracy_score(labels, predictions) * 100))
print("Precision Score : {:.2f} %".format(precision_score(labels, predictions, average='print("Recall Score : {:.2f} %".format(recall_score(labels, predictions, average='mic
```

Train Accuracy : 98.29 %
Test Accuracy : 96.77 %
Precision Score : 96.77 %
Recall Score : 96.77 %

Confusion Matrix

In [9]:

<Figure size 1440x360 with 0 Axes>



In []: