Machine learning with Python-ECL443 Major Project

Problem statement

Classification of plant leaves into their plant species and whether healthy or diseased. A real time prediction using IP Webcam Mobile App was devised.

Plants Considered:

- Guava
- Pomegranate
- Lemon
- Mango

Data collection for testing was done from the premises of boy's hostel canteen.

Model Architecture

Convolutional Layers: These layers extract features from images, and they become deeper and wider as you go deeper into the network.

Depthwise Separable Convolutions: EfficientNet uses efficient depthwise separable convolutions to reduce parameters and computations, consisting of depthwise convolutions.

Squeeze-and-Excitation Blocks: These blocks help the model focus on important features by adaptively recalibrating channel-wise responses.

Inverted Residual Blocks: These blocks introduce non-linearity and feature reuse for better representation without adding much complexity.

Widths, Depths, and Resolution: The model's "B3" designation indicates its specific size, determined by scaling depth, width, and resolution from a baseline architecture.

Resolution Scaling: EfficientNet can handle different image sizes without retraining, making it versatile for various tasks.

Efficient Scaling: EfficientNet balances model size and computational cost using a compound scaling method, achieving good performance across a range of tasks.

Training Dataset













number of classes in the processed dataset= 8 the maximum files in any class in train_df is 201 the minimum files in any class in train_df is 54 train_df length: 1154 test_df length: 248 valid_df length: 247

Collected Testing Dataset:

Lemon (P10)-diseased



















Collected Dataset:























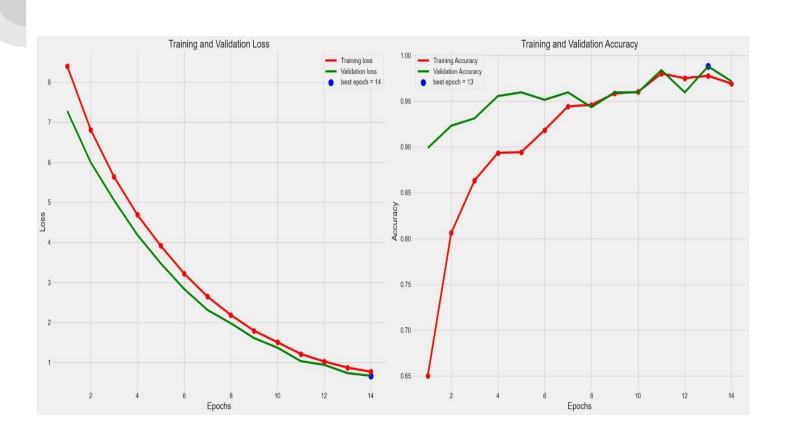




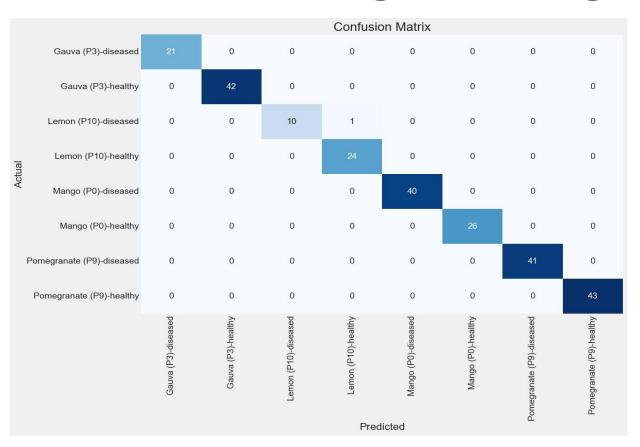




Training & Validation plots



Confusion Matrix on given testing data:

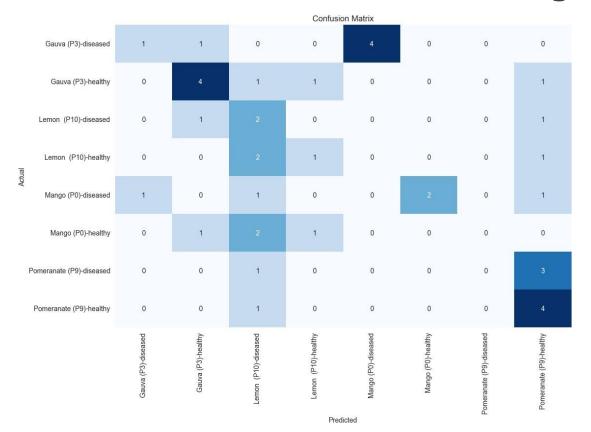


Report for testing of given Testing dataset:

C	l	a	S	S	i	f	i	C	a	t	i	0	n		R	e	p	0	r	t	:
-	_	-	_	_	_	_	-	_	_	_	_	_	_	_	_	_	_	_	_	_	-

	precision	recall	f1-score	support
Gauva (P3)-diseased	1.0000	1.0000	1.0000	21
Gauva (P3)-healthy	1.0000	1.0000	1.0000	42
Lemon (P10)-diseased	1.0000	0.9091	0.9524	11
Lemon (P10)-healthy	0.9600	1.0000	0.9796	24
Mango (P0)-diseased	1.0000	1.0000	1.0000	40
Mango (P0)-healthy	1.0000	1.0000	1.0000	26
Pomegranate (P9)-diseased	1.0000	1.0000	1.0000	41
Pomegranate (P9)-healthy	1.0000	1.0000	1.0000	43
accuracy			0.9960	248
macro avg	0.9950	0.9886	0.9915	248
weighted avg	0.9961	0.9960	0.9959	248

Confusion Matrix on collected testing data:



Report for testing of collected Testing dataset:

	precision	recall	f1-score	support
Gauva (P3)-diseased	0.5000	0.1667	0.2500	6
Gauva (P3)-healthy	0.5714	0.5714	0.5714	7
Lemon (P10)-diseased	0.2000	0.5000	0.2857	4
Lemon (P10)-healthy	0.3333	0.2500	0.2857	4
Mango (P0)-diseased	0.0000	0.0000	0.0000	5
Mango (P0)-healthy	0.0000	0.0000	0.0000	4
Pomeranate (P9)-diseased	0.0000	0.0000	0.0000	4
Pomeranate (P9)-healthy	0.3636	0.8000	0.5000	5
accuracy			0.3077	39
macro avg	0.2460	0.2860	0.2366	39
weighted avg	0.2808	0.3077	0.2637	39

Real Time Prediction:



