



PROBLEM STATEMENT

- Comparative analysis to find the most optimized method in finding the change in multi temporal satellite images.
- Classification of satellite images into a desired labeled dataset.



Model training-(Classification) (Supervised Learning) ResNet50 Model

```
Epoch 1/10
101/101 [=====] - 2469s 24s/step - loss: 1.6674 - accuracy: 0.6438 - val_loss: 0.9265 - val_accuracy: 0.6335
Epoch 2/10
101/101 [=====] - 631s 6s/step - loss: 0.9933 - accuracy: 0.6692 - val_loss: 1.0030 - val_accuracy: 0.6907
Epoch 3/10
101/101 [=====] - 626s 6s/step - loss: 0.7909 - accuracy: 0.6962 - val_loss: 0.8407 - val_accuracy: 0.6683
Epoch 4/10
101/101 [=====] - 621s 6s/step - loss: 0.7274 - accuracy: 0.7068 - val_loss: 1.0872 - val_accuracy: 0.6994
Epoch 5/10
101/101 [=====] - 605s 6s/step - loss: 0.6406 - accuracy: 0.7316 - val_loss: 0.6237 - val_accuracy: 0.7826
Epoch 6/10
101/101 [=====] - 611s 6s/step - loss: 0.7361 - accuracy: 0.7148 - val_loss: 0.7497 - val_accuracy: 0.6211
Epoch 7/10
101/101 [=====] - 616s 6s/step - loss: 0.8750 - accuracy: 0.6922 - val_loss: 0.7452 - val_accuracy: 0.7031
Epoch 8/10
101/101 [=====] - 595s 6s/step - loss: 0.7010 - accuracy: 0.7154 - val_loss: 0.9345 - val_accuracy: 0.7665
Epoch 9/10
101/101 [=====] - 591s 6s/step - loss: 0.6692 - accuracy: 0.7393 - val_loss: 0.8542 - val_accuracy: 0.7043
Epoch 10/10
101/101 [=====] - 610s 6s/step - loss: 0.6987 - accuracy: 0.7278 - val_loss: 0.4850 - val_accuracy: 0.7727
```

ResNet50 model with optimiser(adam)

```
history = model.fit(train_dataset, validation_data=test_dataset, epochs=10, batch_size=32)
```

```
Epoch 1/10
101/101 [=====] - 632s 6s/step - loss: 1.2980 - accuracy: 0.6631 - val_loss: 1.5225 - val_accuracy: 0.6484
Epoch 2/10
101/101 [=====] - 351s 3s/step - loss: 0.8470 - accuracy: 0.6913 - val_loss: 0.9988 - val_accuracy: 0.7006
Epoch 3/10
101/101 [=====] - 351s 3s/step - loss: 0.6660 - accuracy: 0.7387 - val_loss: 0.6026 - val_accuracy: 0.7739
Epoch 4/10
101/101 [=====] - 350s 3s/step - loss: 0.6367 - accuracy: 0.7418 - val_loss: 0.8202 - val_accuracy: 0.6994
Epoch 5/10
101/101 [=====] - 349s 3s/step - loss: 0.6299 - accuracy: 0.7443 - val_loss: 0.5192 - val_accuracy: 0.8050
Epoch 6/10
101/101 [=====] - 352s 3s/step - loss: 0.5493 - accuracy: 0.7743 - val_loss: 0.4986 - val_accuracy: 0.7553
Epoch 7/10
101/101 [=====] - 353s 3s/step - loss: 0.5506 - accuracy: 0.7855 - val_loss: 0.7973 - val_accuracy: 0.7776
Epoch 8/10
101/101 [=====] - 352s 3s/step - loss: 0.5611 - accuracy: 0.7932 - val_loss: 0.6385 - val_accuracy: 0.7565
Epoch 9/10
101/101 [=====] - 357s 4s/step - loss: 0.4634 - accuracy: 0.8180 - val_loss: 0.5272 - val_accuracy: 0.7665
Epoch 10/10
101/101 [=====] - 352s 3s/step - loss: 0.4975 - accuracy: 0.8149 - val_loss: 0.5434 - val_accuracy: 0.7901
```

DenseNet model (without optimiser)

```
print("Total time taken by the model:", total_time, "seconds")
```

```
Epoch 1/10
101/101 [=====] - 2688s 26s/step - loss: 0.7034 - accuracy: 0.7415 - val_loss: 0.4657 - val_accuracy: 0.8211
Epoch 2/10
101/101 [=====] - 11s 106ms/step - loss: 0.4126 - accuracy: 0.8149 - val_loss: 0.4958 - val_accuracy: 0.8224
Epoch 3/10
101/101 [=====] - 11s 105ms/step - loss: 0.3202 - accuracy: 0.8571 - val_loss: 0.6496 - val_accuracy: 0.7826
Epoch 4/10
101/101 [=====] - 11s 106ms/step - loss: 0.2427 - accuracy: 0.9014 - val_loss: 0.9986 - val_accuracy: 0.7739
Epoch 5/10
101/101 [=====] - 10s 102ms/step - loss: 0.1984 - accuracy: 0.9194 - val_loss: 0.9814 - val_accuracy: 0.7540
Epoch 6/10
101/101 [=====] - 11s 109ms/step - loss: 0.1550 - accuracy: 0.9402 - val_loss: 1.4511 - val_accuracy: 0.7466
Epoch 7/10
101/101 [=====] - 11s 106ms/step - loss: 0.1294 - accuracy: 0.9507 - val_loss: 1.0420 - val_accuracy: 0.7354
Epoch 8/10
101/101 [=====] - 11s 106ms/step - loss: 0.1147 - accuracy: 0.9616 - val_loss: 1.2843 - val_accuracy: 0.7540
Epoch 9/10
101/101 [=====] - 11s 105ms/step - loss: 0.0930 - accuracy: 0.9671 - val_loss: 1.8285 - val_accuracy: 0.7217
Epoch 10/10
101/101 [=====] - 11s 105ms/step - loss: 0.0855 - accuracy: 0.9699 - val_loss: 1.9735 - val_accuracy: 0.6969
Total time taken by the model: 2838.1581342220306 seconds
```

Completed at 2:10 PM

DenseNet model (with optimiser)

```
Epoch 1/10
101/101 [=====] - 802s 8s/step - loss: 0.5626 - accuracy: 0.7616 - val_loss: 0.4111 - val_accuracy: 0.8037
Epoch 2/10
101/101 [=====] - 11s 105ms/step - loss: 0.3744 - accuracy: 0.8292 - val_loss: 0.4154 - val_accuracy: 0.8273
Epoch 3/10
101/101 [=====] - 10s 101ms/step - loss: 0.2885 - accuracy: 0.8723 - val_loss: 0.5264 - val_accuracy: 0.7925
Epoch 4/10
101/101 [=====] - 10s 103ms/step - loss: 0.2233 - accuracy: 0.9086 - val_loss: 0.8865 - val_accuracy: 0.7627
Epoch 5/10
101/101 [=====] - 10s 100ms/step - loss: 0.1777 - accuracy: 0.9275 - val_loss: 0.6794 - val_accuracy: 0.7590
Epoch 6/10
101/101 [=====] - 10s 99ms/step - loss: 0.1455 - accuracy: 0.9473 - val_loss: 0.8133 - val_accuracy: 0.7379
Epoch 7/10
101/101 [=====] - 10s 100ms/step - loss: 0.1155 - accuracy: 0.9551 - val_loss: 1.0163 - val_accuracy: 0.7578
Epoch 8/10
101/101 [=====] - 10s 100ms/step - loss: 0.0818 - accuracy: 0.9709 - val_loss: 1.0706 - val_accuracy: 0.7839
Epoch 9/10
101/101 [=====] - 11s 104ms/step - loss: 0.0768 - accuracy: 0.9715 - val_loss: 1.2101 - val_accuracy: 0.6832
Epoch 10/10
101/101 [=====] - 10s 100ms/step - loss: 0.0690 - accuracy: 0.9771 - val_loss: 1.2956 - val_accuracy: 0.7031
Total time taken by the model: 912.7276697158813 seconds
```

Inception resnet (without optimizer)

Epoch 1/10

102/102 [=====] - 1310s 13s/step - loss: 1.5030 - accuracy: 0.7010 - val_loss: 0.4473 - val_accuracy: 0.8463

Epoch 2/10

102/102 [=====] - 22s 218ms/step - loss: 0.4986 - accuracy: 0.7891 - val_loss: 0.4059 - val_accuracy: 0.8315

Epoch 3/10

102/102 [=====] - 22s 213ms/step - loss: 0.3588 - accuracy: 0.8476 - val_loss: 0.5111 - val_accuracy: 0.8265

Epoch 4/10

102/102 [=====] - 22s 215ms/step - loss: 0.2808 - accuracy: 0.8878 - val_loss: 0.5334 - val_accuracy: 0.8030

Epoch 5/10

102/102 [=====] - 23s 221ms/step - loss: 0.2053 - accuracy: 0.9246 - val_loss: 0.7353 - val_accuracy: 0.7819

Epoch 6/10

102/102 [=====] - 24s 232ms/step - loss: 0.1765 - accuracy: 0.9419 - val_loss: 0.8133 - val_accuracy: 0.7807

Epoch 7/10

102/102 [=====] - 23s 220ms/step - loss: 0.1416 - accuracy: 0.9539 - val_loss: 0.9409 - val_accuracy: 0.7943

Epoch 8/10

102/102 [=====] - 23s 220ms/step - loss: 0.1268 - accuracy: 0.9576 - val_loss: 1.2106 - val_accuracy: 0.7794

Epoch 9/10

102/102 [=====] - 23s 228ms/step - loss: 0.1100 - accuracy: 0.9666 - val_loss: 1.1047 - val_accuracy: 0.7968

Epoch 10/10

102/102 [=====] - 23s 220ms/step - loss: 0.0955 - accuracy: 0.9706 - val_loss: 1.4298 - val_accuracy: 0.7770

Total time taken by the model: 1562.1777930259705 seconds

Inception resnet(with optimizer)

Epoch 1/10

102/102 [=====] - 40s 254ms/step - loss: 1.0386 - accuracy: 0.7066 - val_loss: 0.5526 - val_accuracy: 0.7175

Epoch 2/10

102/102 [=====] - 23s 221ms/step - loss: 0.4593 - accuracy: 0.7823 - val_loss: 0.4629 - val_accuracy: 0.8228

Epoch 3/10

102/102 [=====] - 22s 216ms/step - loss: 0.3490 - accuracy: 0.8550 - val_loss: 0.6563 - val_accuracy: 0.7559

Epoch 4/10

102/102 [=====] - 23s 225ms/step - loss: 0.2590 - accuracy: 0.8949 - val_loss: 0.5849 - val_accuracy: 0.7745

Epoch 5/10

102/102 [=====] - 24s 235ms/step - loss: 0.2116 - accuracy: 0.9184 - val_loss: 0.6715 - val_accuracy: 0.7770

Epoch 6/10

102/102 [=====] - 22s 217ms/step - loss: 0.2261 - accuracy: 0.9159 - val_loss: 0.7235 - val_accuracy: 0.7869

Epoch 7/10

102/102 [=====] - 23s 222ms/step - loss: 0.1497 - accuracy: 0.9437 - val_loss: 0.8575 - val_accuracy: 0.7596

Epoch 8/10

102/102 [=====] - 23s 227ms/step - loss: 0.1302 - accuracy: 0.9601 - val_loss: 1.0228 - val_accuracy: 0.7633

Epoch 9/10

102/102 [=====] - 23s 220ms/step - loss: 0.0814 - accuracy: 0.9722 - val_loss: 1.0612 - val_accuracy: 0.7497

Epoch 10/10

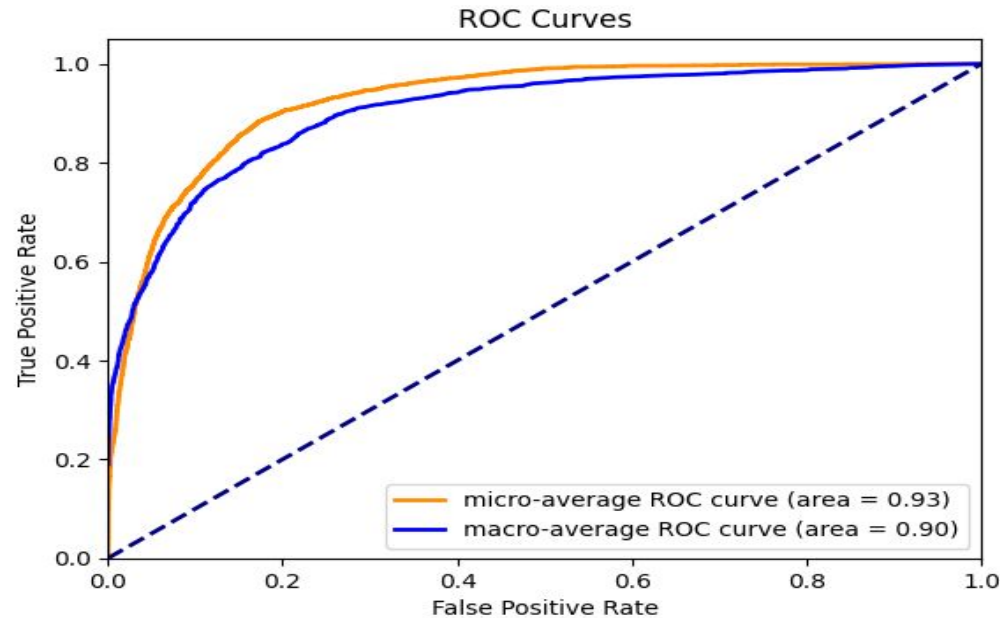
102/102 [=====] - 23s 222ms/step - loss: 0.1027 - accuracy: 0.9647 - val_loss: 1.1000 - val_accuracy: 0.7869

Total time taken by the model: 245.01757049560547 seconds

ResNet50 model metrics

	Without Optimizer	With Optimizer(adam)
Accuracy	0.764	0.801
Precision Score	0.900	0.899
Recall Score	0.856	0.871
F1 Score	0.865	0.876
Error Rate	0.235	0.218
Computational Time(in sec)	8121.742	6101.651

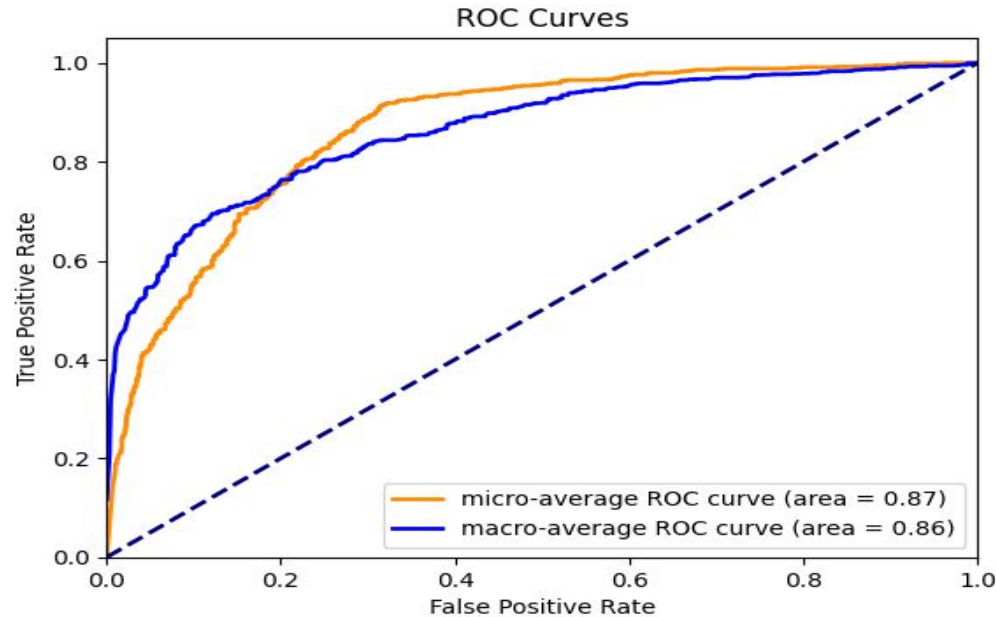
ROC Curve and AUC(ResNet50)



DenseNet model metrics

	Without Optimizer	With Optimizer(adam)
Accuracy	0.699	0.720
Precision Score	0.718	0.871
Recall Score	0.776	0.843
F1 Score	0.820	0.836
Error Rate	0.308	0.300
Computational Time(in secs)	2838.158	912.727 seconds

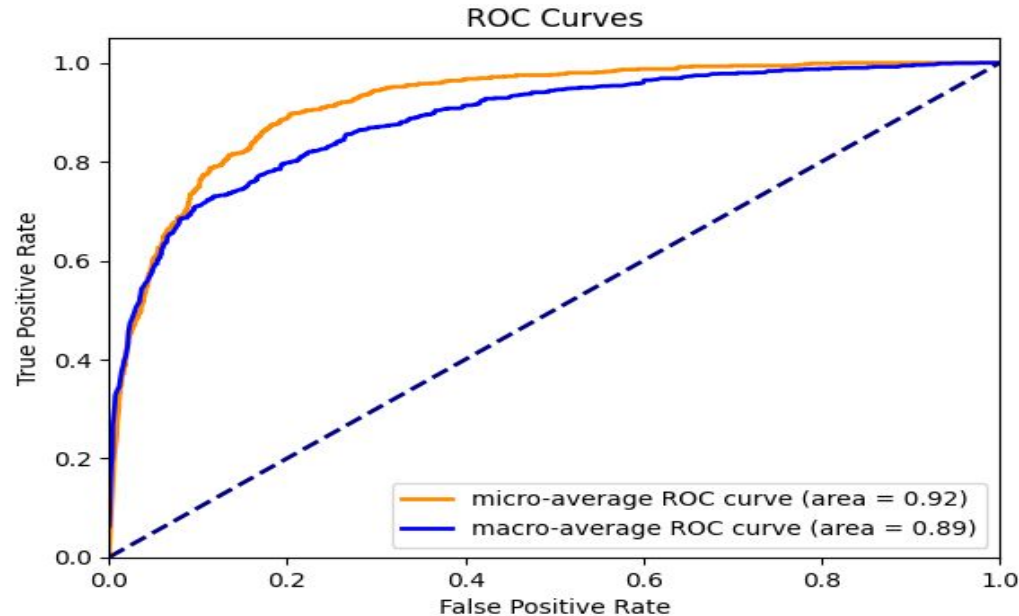
ROC Curve and AUC(DenseNet)



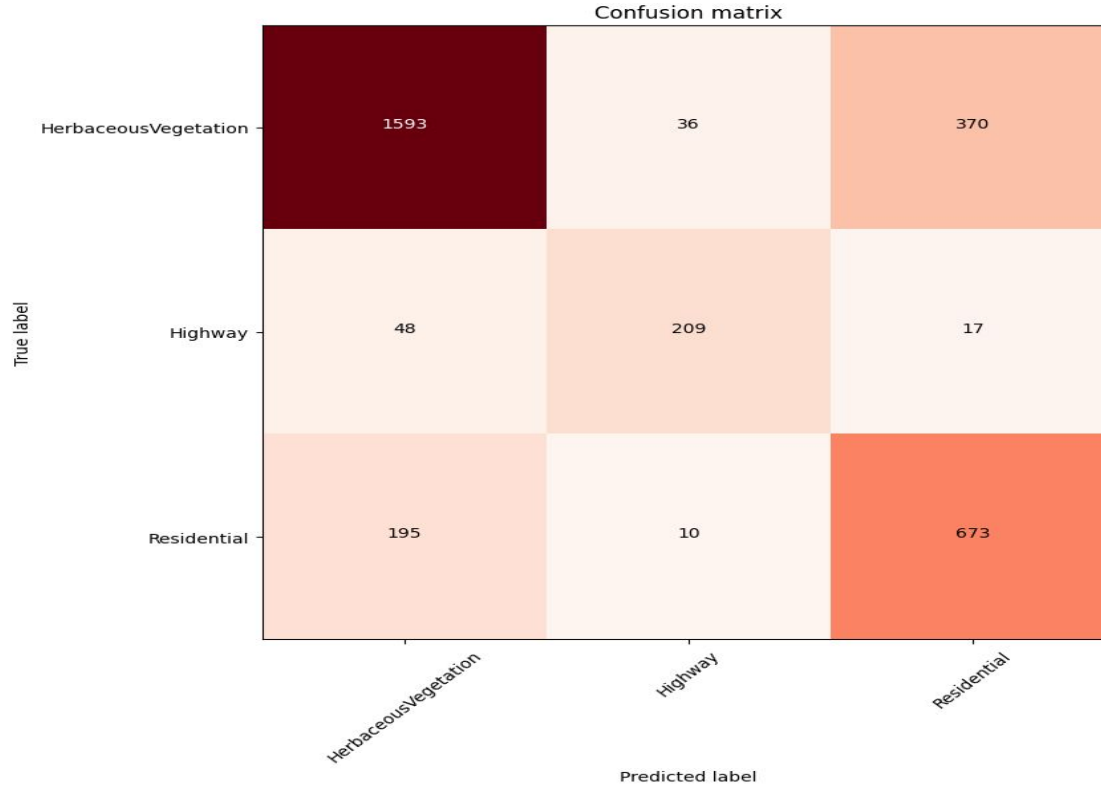
Inception resnet model metrics

	Without Optimizer	With Optimizer(adam)
Accuracy	0.777	0.780
Precision Score	0.842	0.877
Recall Score	0.828	0.895
F1 Score	0.865	0.872
Error Rate	0.277	0.223
Computational Time(in sec)	1562.177	245.017

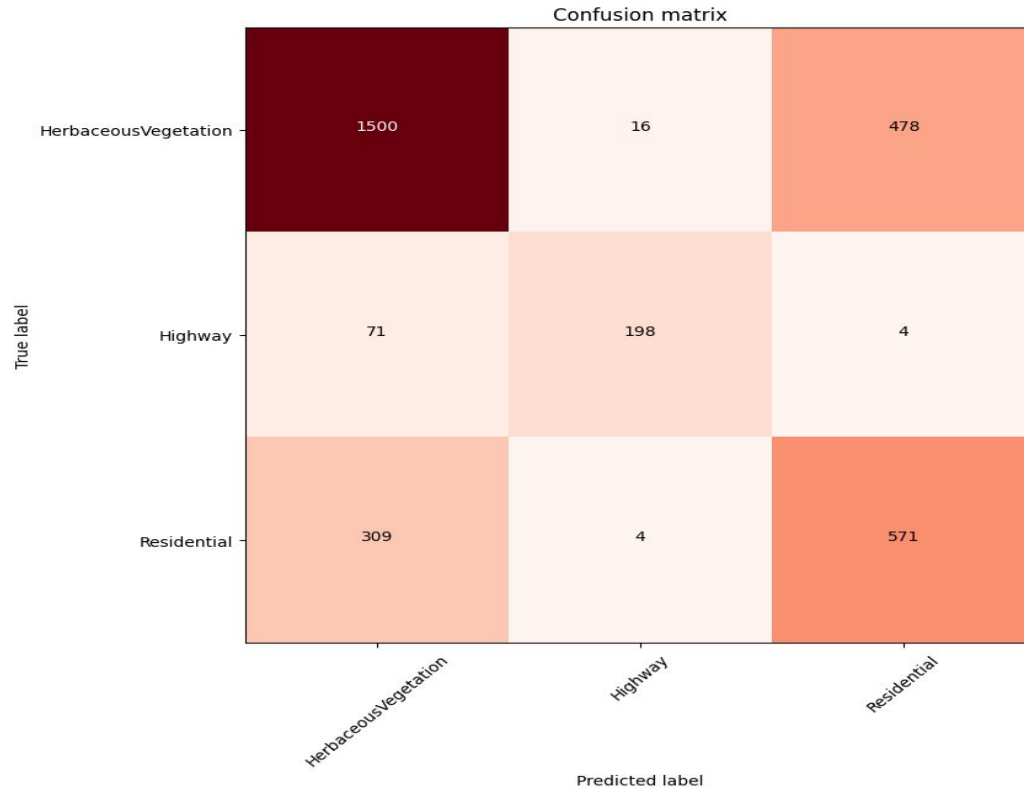
ROC Curve and AUC(Inception resnet)



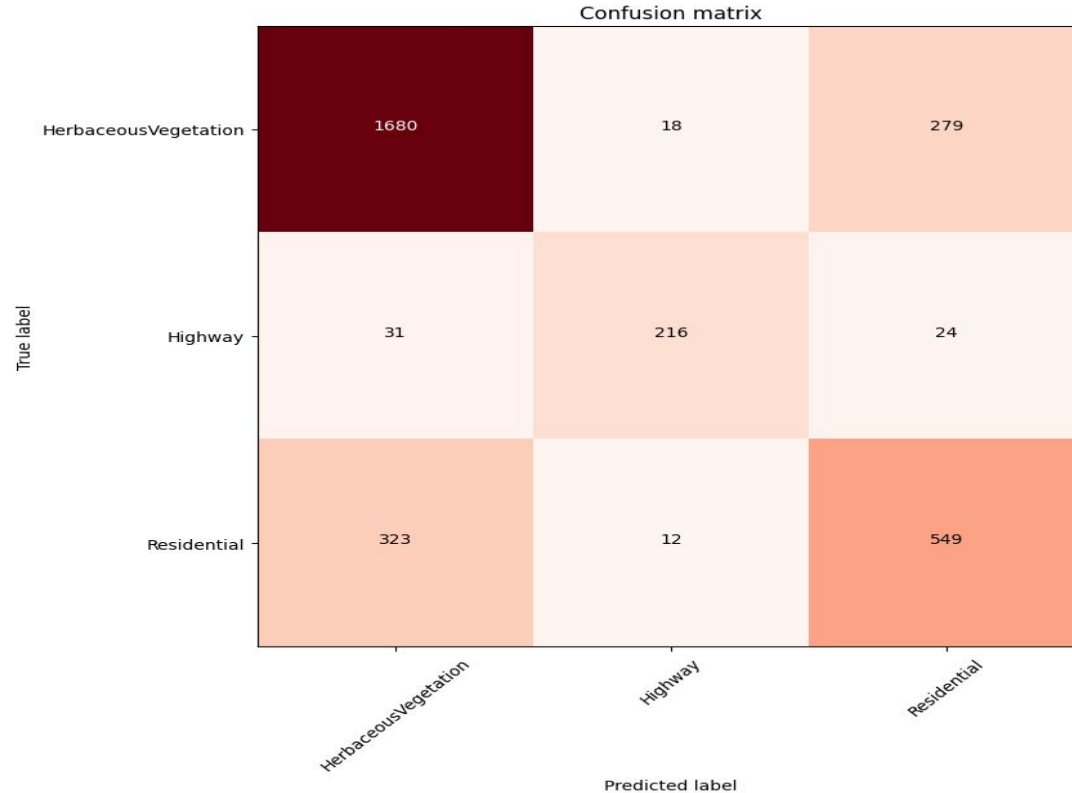
Confusion matrix (ResNet50)



Confusion matrix (DenseNet)



Confusion matrix (Inception)



RESULT ANALYSIS

- The accuracies of 4 different clustering methods(K-Means, Mean Shift, DT-CWT, PSO) are compared using a table consisting of Pfa, Pma Pte as comparison metrics.
- In image clustering, the clustering method with least metric value is considered the most effective.
- In image classification, the input dataset is fed into a deep learning algorithm(ResNet model) for better effective and efficient classification.
- The accuracy at each epoch is varied and the final epoch's(10) accuracy of 81% is achieved.
- The accuracy follows an increasing trend when plotted against epoch.
- The loss over each epoch is significantly decreasing with increase in epoch.
- In the confusion matrix, the true positives(diagonal elements) are high implying the overall accuracy is maximum.

CONCLUSIONS

- From the histogram the PSO method has the least Pfa, Pte and Pma values, implying that it is the most effective method.
- In image classification, the accuracy is optimal considering the dataset size and number of parameters.
- The final change mask can also be used for image segmentation in order to quantify the individual changes.
- The classification output can be used in calculating land cover land change over the years.