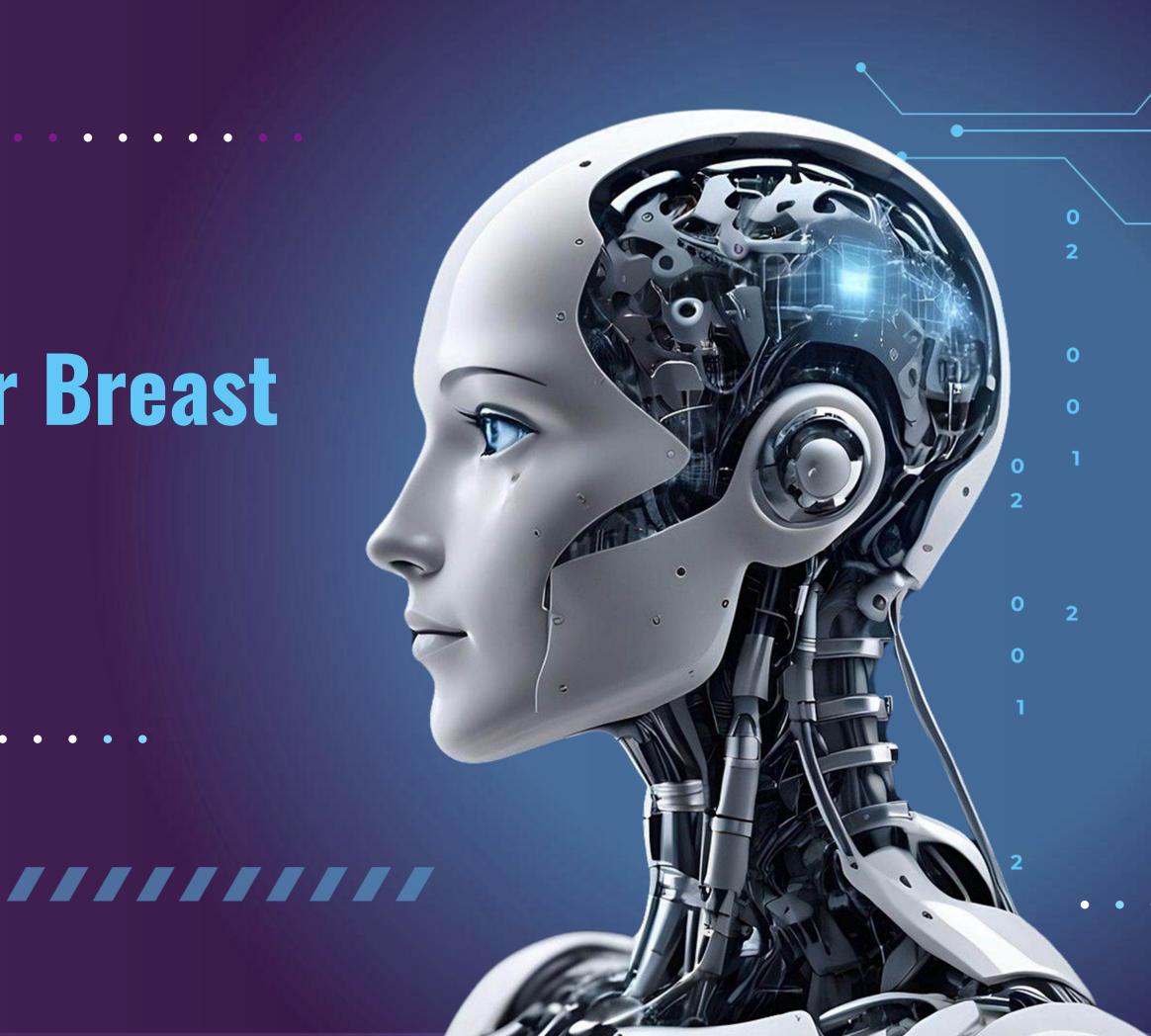


TUMOR TRACE: MRI – Based AI for Breast Cancer Detection



BY MU DARSHAN





PROJECT OVERVIEW

Objective:

Build a machine learning model to classify breast cancer status (benign or malignant) using MRI images.

Goal:

Aid in the early detection of breast cancer to improve patient outcomes.

Approach:

Utilize deep learning models, data preprocessing, and evaluation metrics.





Problem Statement!!

The Problem:

Breast cancer is a leading cause of death among women worldwide.

Early detection plays a vital role in improving survival rates.

Manual analysis of MRI scans is slow, expensive, and prone to errors.

The Need:

Automated, accurate, and fast classification of MRI scans to detect cancer earlier and more reliably.

This project aims to develop a ml model that can accurately classify MRI images as benign or malignant there by aiding in early dectection and diagonosis of breast cancer

Project Goal:

Develop a machine learning model that can classify MRI images as benign or malignant, aiding in early detection and diagnosis.







Data collection from MRI images. Preprocessing for model input. Model training and evaluation.



Data Collection - Obtain labeled MRI datasets.

Data Preprocessing - Resize images, normalize pixel values, data augmentation.

Model Development - Implement deep learning models (VGG16, ResNet).

Evaluation - Assess performance using metrics like accuracy, precision, recall, F1 score.

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Data Collection and Preprocessing



Dataset:

Contains MRI images labeled as either benign or malignant.

Source of Dataset:

https://www.sciencedirect.com/science/article/abs/pii/S00104

82523007205



Preprocessing Steps:

Grayscale Conversion: Convert images to grayscale.

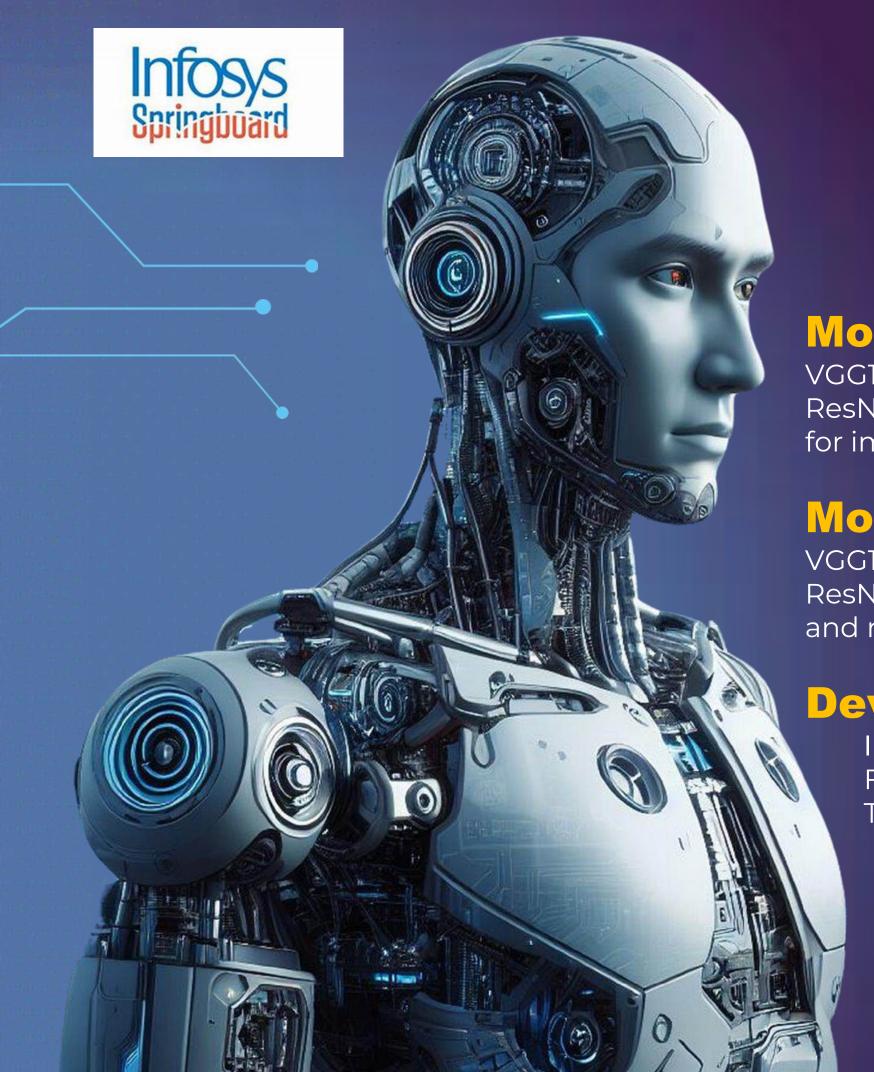
Resizing: Resize images to a standard dimension (e.g., 224x224).

Normalization: Normalize pixel values to a range suitable for neural network input.

Augmentation: Apply techniques like rotation and flipping to augment the dataset.

Handling Class Imbalance: Use techniques like oversampling, undersampling, and

class weights..



Models Used and Development

Models Used:

VGG16: A convolutional neural network with 16 layers.
ResNet18 & ResNet50: Deeper architectures using residual connections for improved training.

Model Selection:

VGG16: Chosen for its simplicity and effectiveness in image classification. ResNet18 & ResNet50: Chosen for better performance with deeper layers and residual learning.

Development Steps:

Initialize pre-trained models.
Fine-tune on the breast cancer dataset.
Train and validate models.



Precision

Ability to avoid false positives.



Ability to identify actual positive cases (sensitivity)...

Recall

F1 Score

Harmonic mean of

precision and recall.



Evaluation Metrics



Accuracy

Percentage of correct predictions.



Confusion Matrix

Helps in understanding the performance of your classification, showing true positives, false positives, etc.

AUC (Area Under Curve)

Measures how well the model distinguishes between benign and malignant tumors.

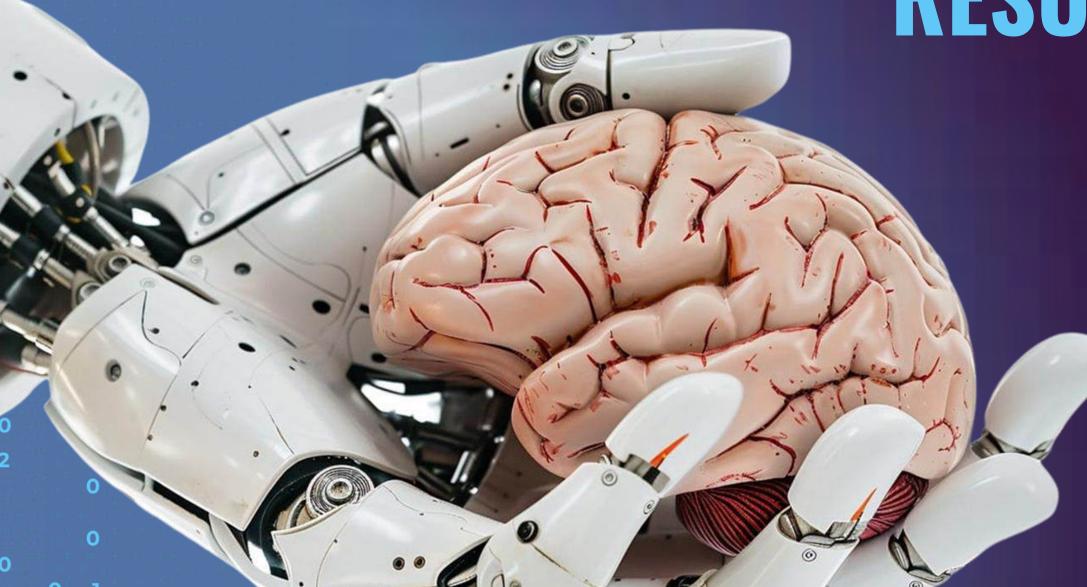
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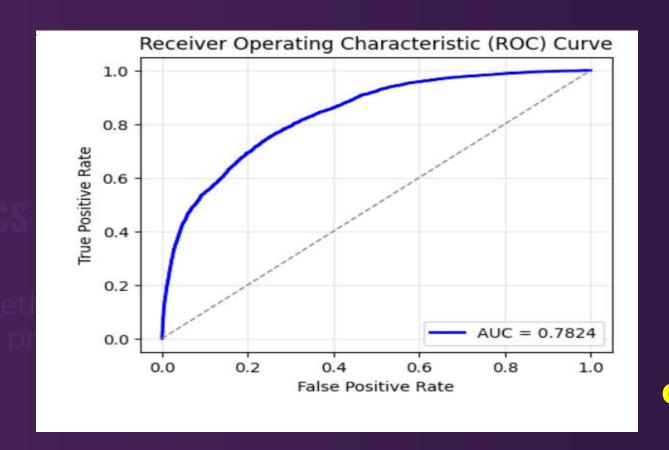


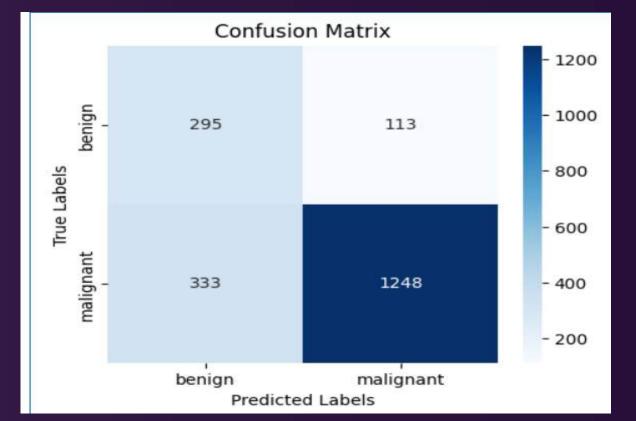
- 1. MODEL VGG16
- 2. MODEL RESNET 18
- 3. MODEL -RESNET 80

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VGG16





CONFUSION MATRIX: [[1005 933] [777 4136]]

recall f1-score precision support benign 0.5640 0.5186 0.5403 1938 malignant 0.8159 0.8418 0.8287 4913 0.7504 6851 accuracy 0.6802 6851 0.6900 0.6845 macro avg weighted avg 0.7504 0.7471 0.7447 6851

SPECIFICITY: 1.0000, SENSITIVITY: 0.0000, AUC: 0.7824

TEST SET:

AVERAGE LOSS: 0.5217,

ACCURACY: 5141/6851 (75.04%)

2

0

0

0

0 1 2

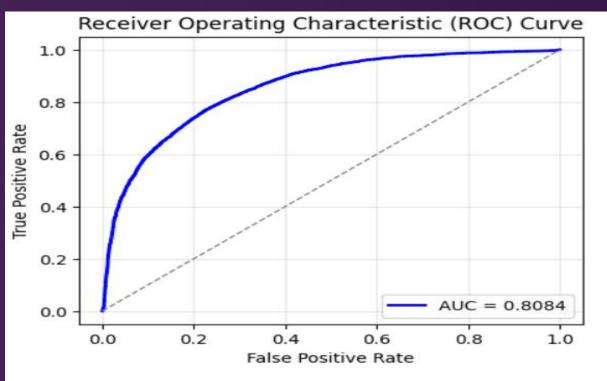
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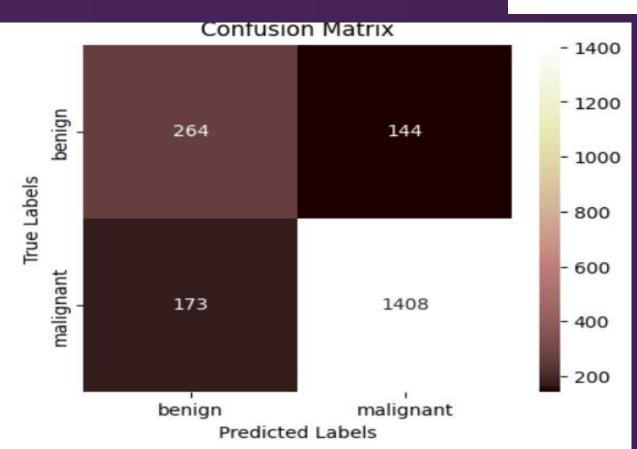
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RESNET 18





	precision	recall	f1-score	support	
benign malignant	0.6338 0.8064	0.4546 0.8964	0.5294 0.8490	1938 4913	
accuracy macro avg weighted avg	0.7201 0.7576	0.6755 0.7714	0.7714 0.6892 0.7586	6851 6851 6851	

Confusion Matrix: [[881 1057] [509 4404]]

Specificity: 1.0000, Sensitivity: 0.0000, AUC: 0.8084

test set:

Average loss: 0.6559

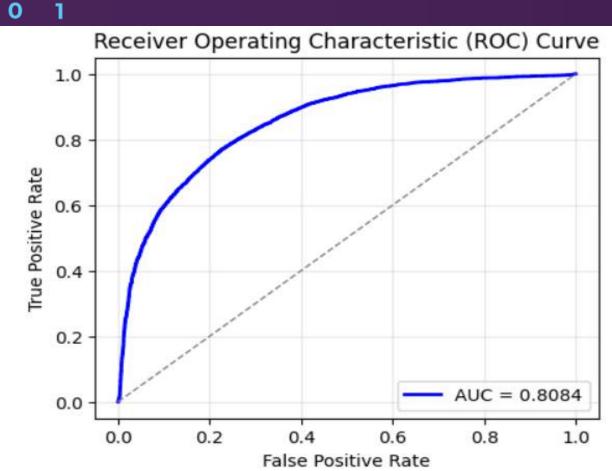
Accuracy: 5285/6851 (77.14%)

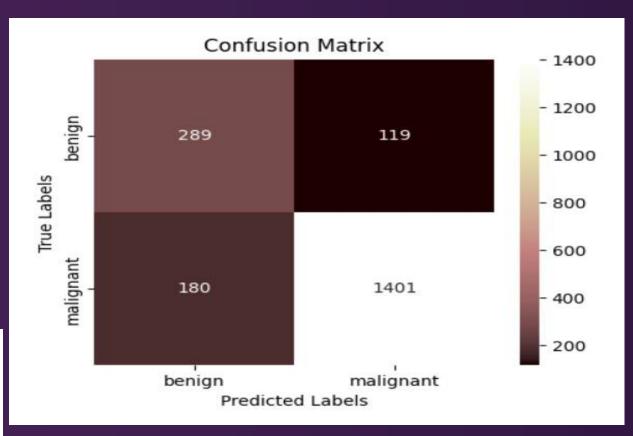
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RESNET80







precision recall f1-score support benign 0.5913 0.4912 0.5366 1938 malignant 0.8119 0.8661 0.8381 4913 accuracy 0.7600 6851 macro avg 0.7016 0.6786 0.6874 6851 weighted avg 0.7495 0.7600 0.7528 6851					
malignant 0.8119 0.8661 0.8381 4913 accuracy 0.7600 6851 macro avg 0.7016 0.6786 0.6874 6851		precision	recall	f1-score	support
accuracy 0.7600 6851 macro avg 0.7016 0.6786 0.6874 6851	benign	0.5913	0.4912	0.5366	1938
macro avg 0.7016 0.6786 0.6874 6851	malignant	0.8119	0.8661	0.8381	4913
macro avg 0.7016 0.6786 0.6874 6851					
	accuracy			0.7600	6851
weighted avg 0.7495 0.7600 0.7528 6851	macro avg	0.7016	0.6786	0.6874	6851
	weighted avg	0.7495	0.7600	0.7528	6851

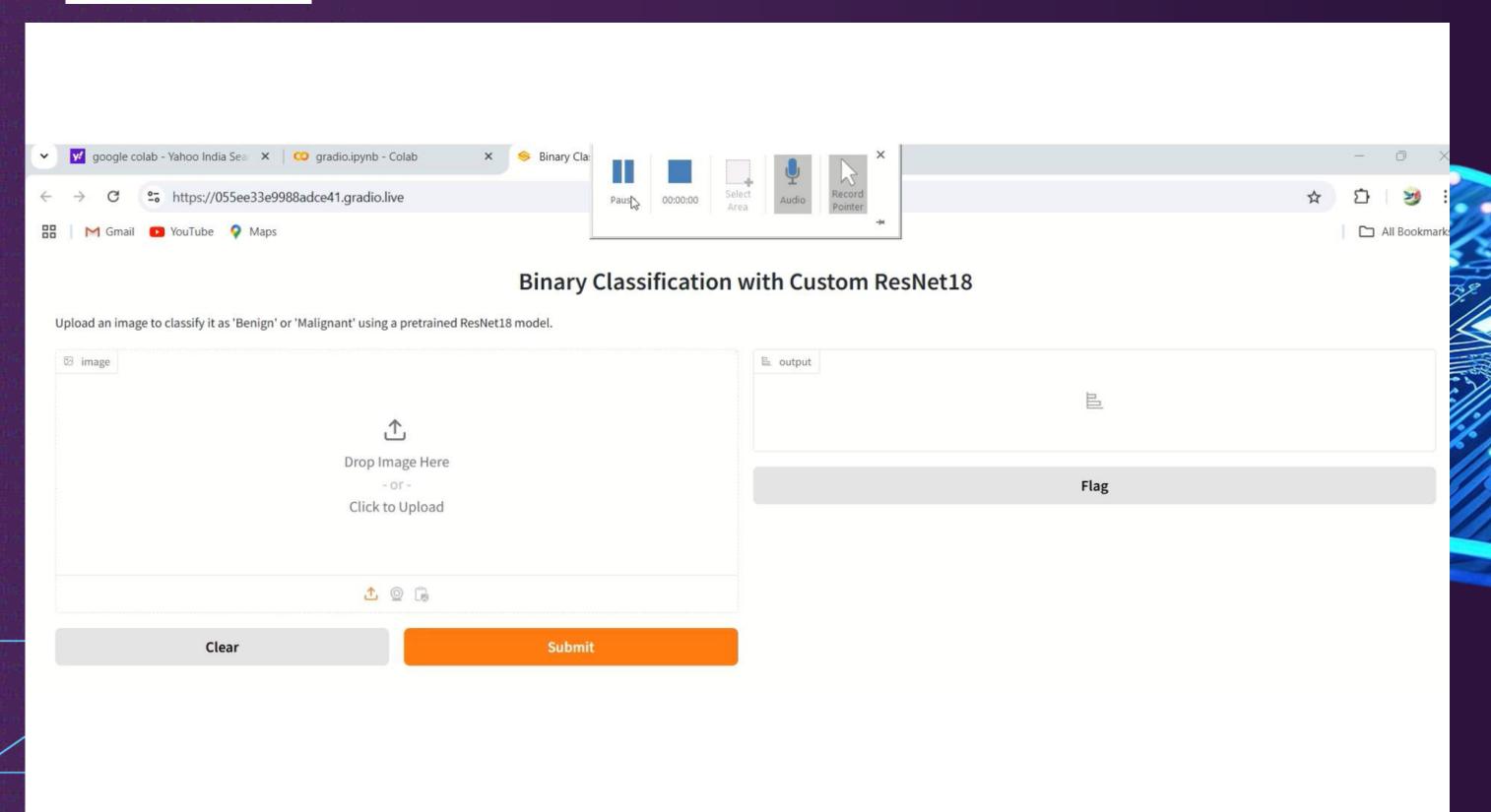
CLIENT Confusion Matrix: [[952 986] [658 4255]]

Specificity: 1.0000, Sensitivity: 0.0000, AUC: 0.7912

test set: Average loss: 0.6254, Accuracy: 5207/6851 (76.00%)



FINAL RESULT





Project: MRI-Based Alfor Breast Cancer

Detection

BY

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