

# **BREAST CANCER CLASSIFICATION AND DETECTION THROUGH ADVANCED PRE-PROCESSING AND DEEP LEARNING TECHNIQUES**

**Group ID: CSE 25-55**

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# COURSE OUTCOMES (COS)

✓ CO2:

Ability to implement image preprocessing techniques for enhancing and preparing medical images for deep learning models.

✓ CO3:

Capability to develop and evaluate deep learning models for classification tasks using appropriate performance metrics.

# PROGRAM OUTCOMES (POS)

## ✓ PO2:

Ability to identify, analyze, and solve complex engineering problems using first principles of mathematics and engineering sciences.

## ✓ PO5:

Proficiency in using modern tools and techniques, such as machine learning frameworks and image processing libraries, for engineering applications.

# SUSTAINABLE DEVELOPMENT GOALS

- ✓ **SDG 3:**  
**Good Health and Well-being** – Early cancer detection improves survival.
- ✓ **SDG 9:**  
**Industry, Innovation, and Infrastructure** – Promotes use of AI in healthcare.
- ✓ **SDG 10:**  
**Reduced Inequality** – Supports access to diagnostic support in rural areas.

# PROJECT ABSTRACTION

- Aim: To design a system for early detection and classification of breast cancer using mammographic images.
- Integrates image preprocessing with deep learning-based classification.
- Enhances diagnostic reliability, reduces false negatives.
- Supports radiologists in clinical decision-making.

# OVERVIEW

- Automates classification of mammograms as benign or malignant.
- Uses preprocessing techniques like CLAHE, edge detection, pectoral muscle removal.
- Trains and compares CNN models: VGG16, ResNet50, MobileNet, and Custom CNN.
- Improves diagnostic speed, precision, and recall.

# KEY OBJECTIVES

- Improve image quality and remove misleading features (e.g., pectoral muscles).
- Apply deep learning to classify tumors.
- Compare model performances to choose the most balanced one.
- Enable scalable and accurate early diagnosis.



# METHODOLOGY

- **Dataset:** Publicly available mammogram datasets (DDSM, INbreast).
- **Preprocessing:** Orientation correction, CLAHE, pectoral muscle removal, morphological operations.
- **Model Training:** CNN-based models trained on binary classification task.
- **Evaluation:** Precision, recall, accuracy, F1-score, confusion matrix.



# MODEL ARCHITECTURE

- **VGG16**: Deep network, good feature extraction, prone to overfitting.
- **ResNet50**: Uses residual blocks to train deeper networks.
- **MobileNet**: Lightweight and efficient for real-time use.
- **Custom CNN**: Built for control and experimentation.

# EVALUATION MATRIX

MODEL	TRAINING ACCURACY	TEST ACCURACY	PRECISION (TEST)	RECALL (TEST)	F1-SCORE (TEST)
VGG16	99.78%	62.26%	63.71%	62.26%	61.50%
ResNet50	97.44%	50.94%	51.16%	50.94%	50.98%
MobileNet	89.96%	60.38%	61.75%	60.38%	59.98%
Custom CNN	99.57%	50.94%	50.74%	50.94%	50.77%

# EXPECTED OUTCOMES

- Successfully developed a breast cancer classification system.
- Achieved high accuracy and model reliability.
- MobileNet identified as best performer.
- Developed a reusable, modular pipeline.

# TESTING AND REFINEMENT

- 80:20 train-test split.
- Data augmentation used to prevent overfitting.
- MobileNet showed the best generalization performance.
- Refinement focused on enhancing preprocessing and model tuning.

# FUTURE SCOPE

- Integrate real-time classification via web/mobile UI.
- Expand model to other medical imaging types (MRI, ultrasound).
- Use explainable AI (e.g., Grad-CAM) for visual justification.
- Train on larger datasets to enhance robustness.

# THANK YOU