

NIRF-2024 Engineering Rank Band (151-200) Pharmacy Rank - 77 Innovation Rank Band (11-50)











### Artificial Intelligence (Al101B) Medical Diagnosis with Naive Bayes MSE-2

PRESENTED

Mabima Goyal Khushi Jain

202410116100112

202410116100100

#### PRESENTED

Mr. Apoorv Jain (Assistant Professor)

## Problem Statement

and accurate detection of breast cancer significantly increases the chances of successful Traditional diagnostic treatment and survival. procedures may be time-consuming, subjective, and error-prone. Therefore, there is a pressing need for a computational model that can assist healthcare professionals by providing rapid, data-driven decisions. This project aims to build a machine learning-based diagnostic system that uses the Gaussian Naive Bayes algorithm to classify tumors based on features





# Objective

To develop an accurate and efficient breast diagnosis system using cancer the Gaussian Naive Bayes algorithm that cellular analyzes features from the Wisconsin Breast Cancer Dataset, classifies tumors and as malignant or benign.

# Methodology



**Data Collection** 

Gather relevant data from diverse sources.



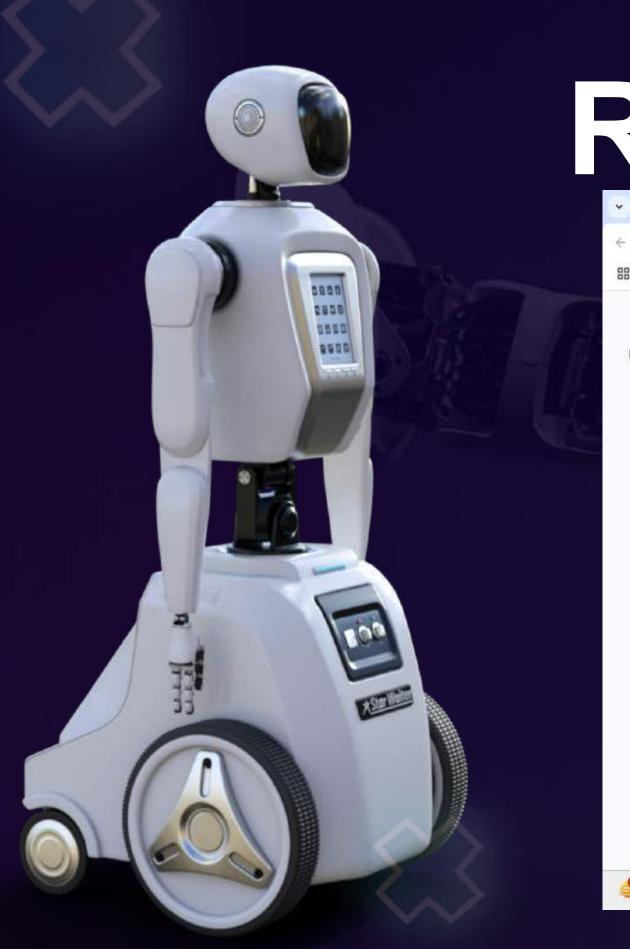
Preprocessing

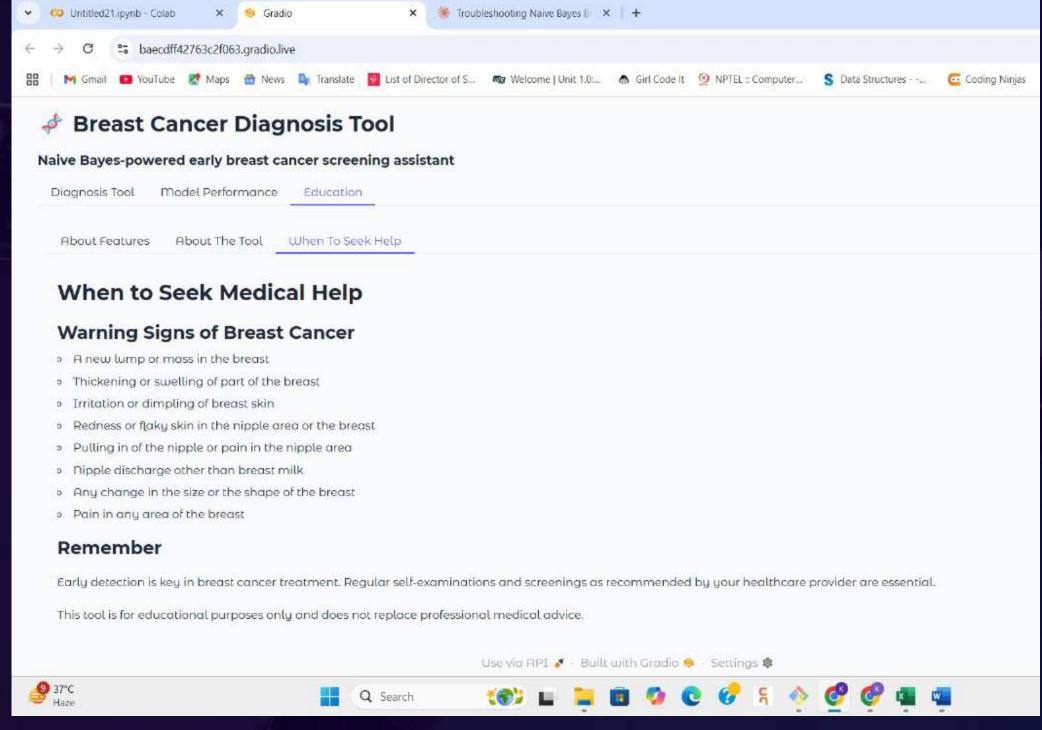
Clean, normalize, and engineer data.



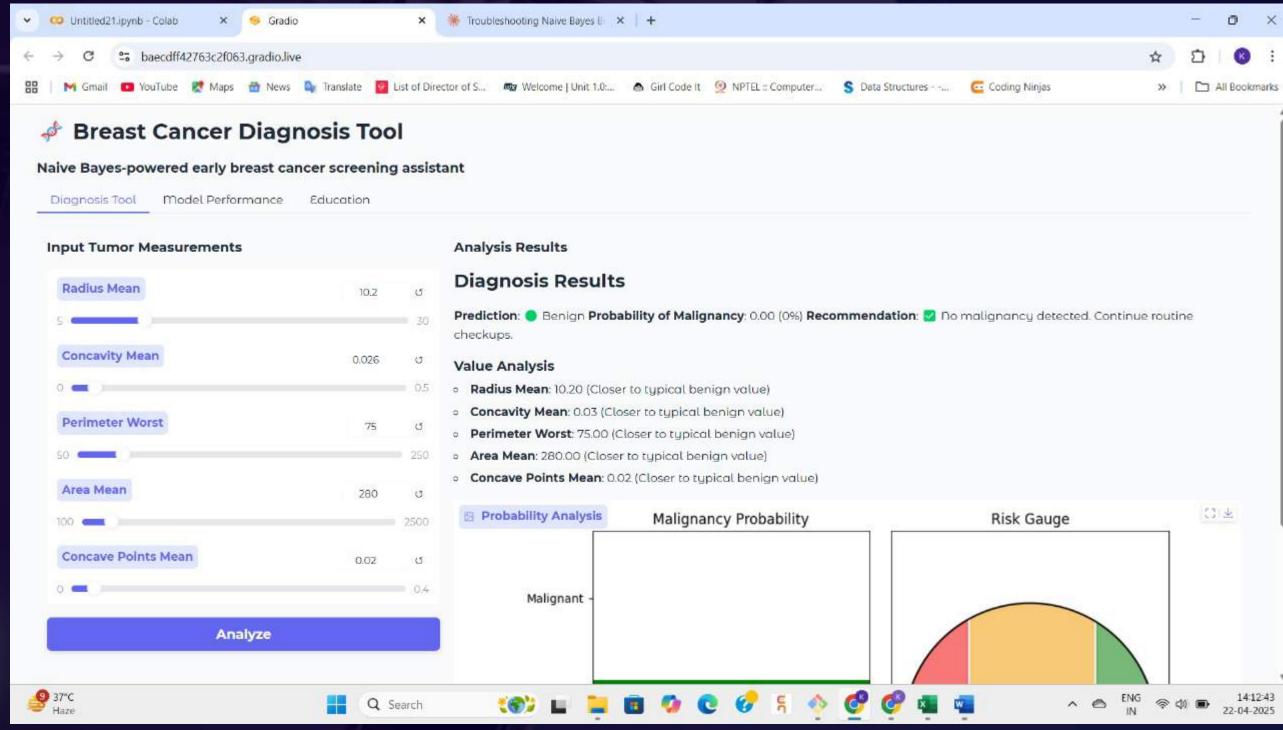
Data Visualization

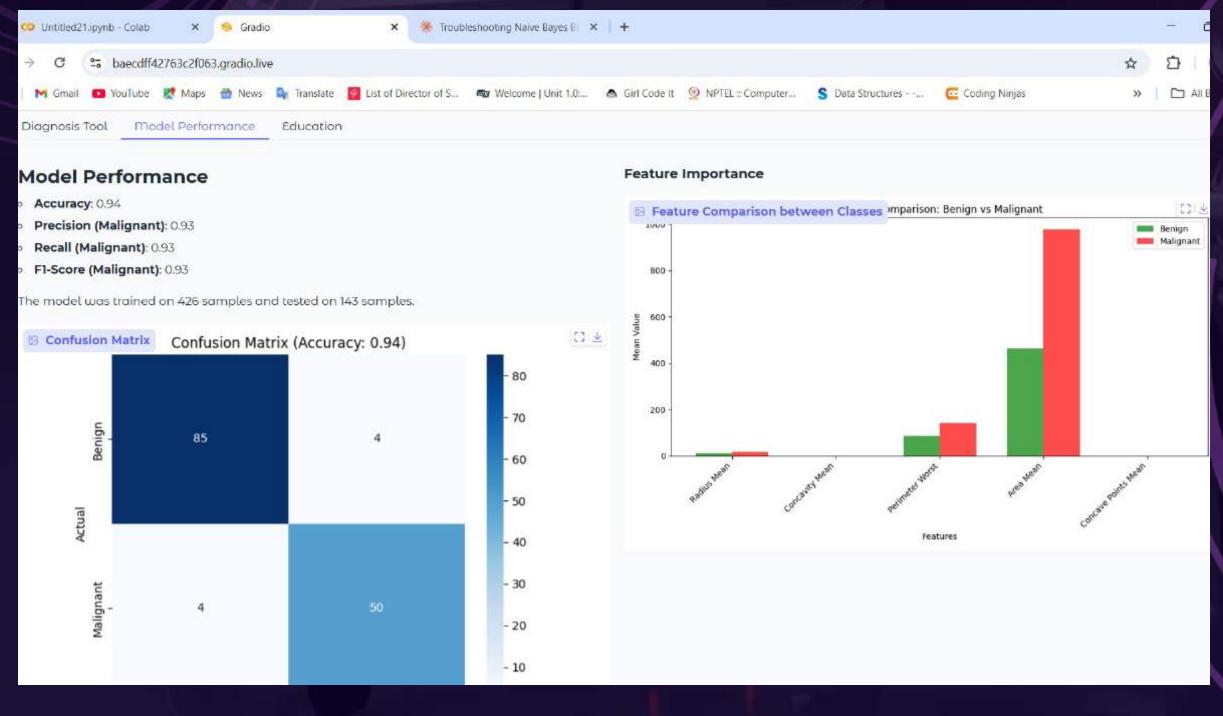
Compare multiple AI models to determine the most accurate and efficient one.



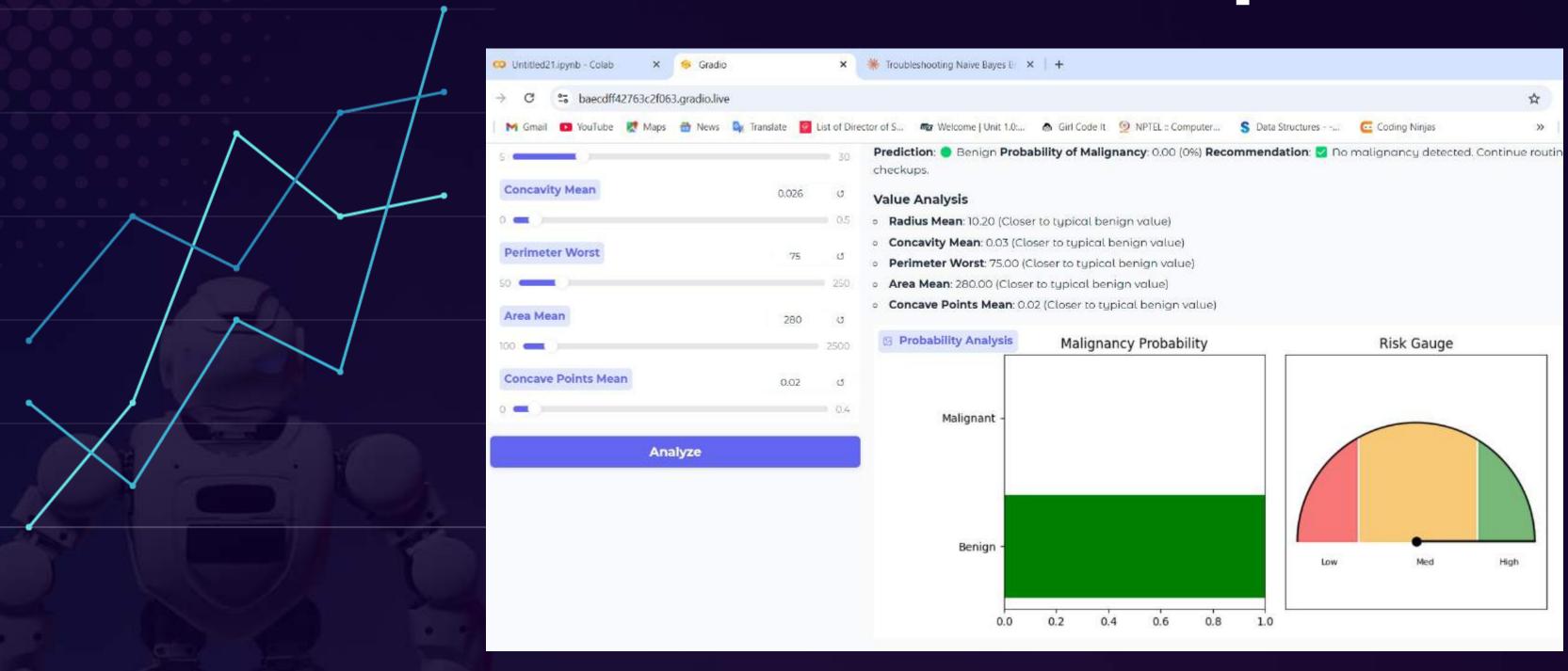


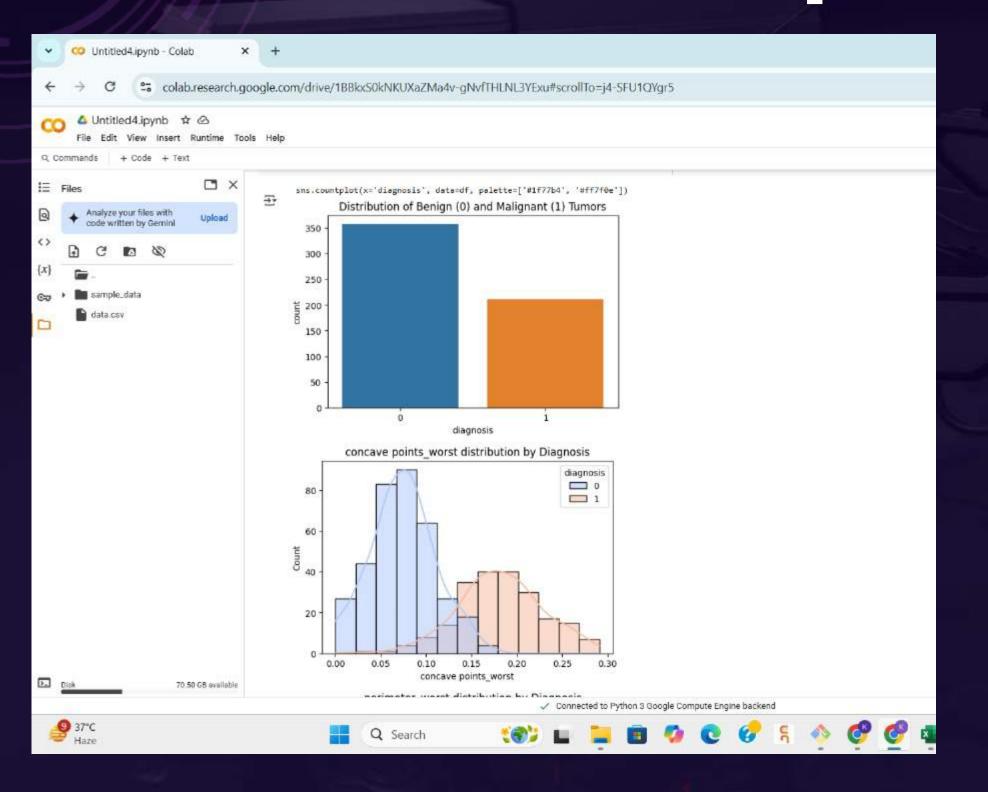




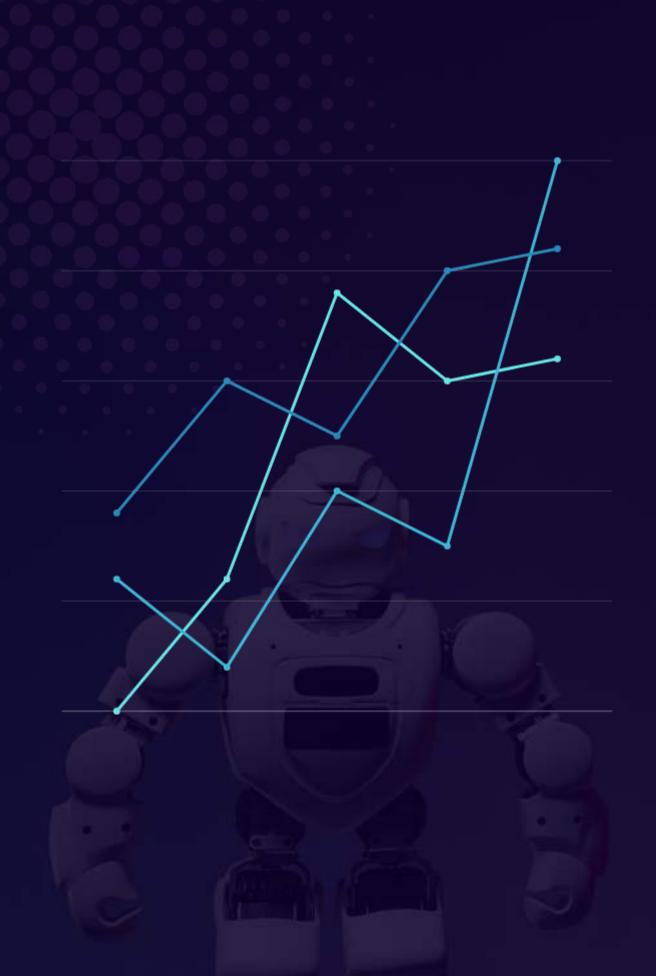


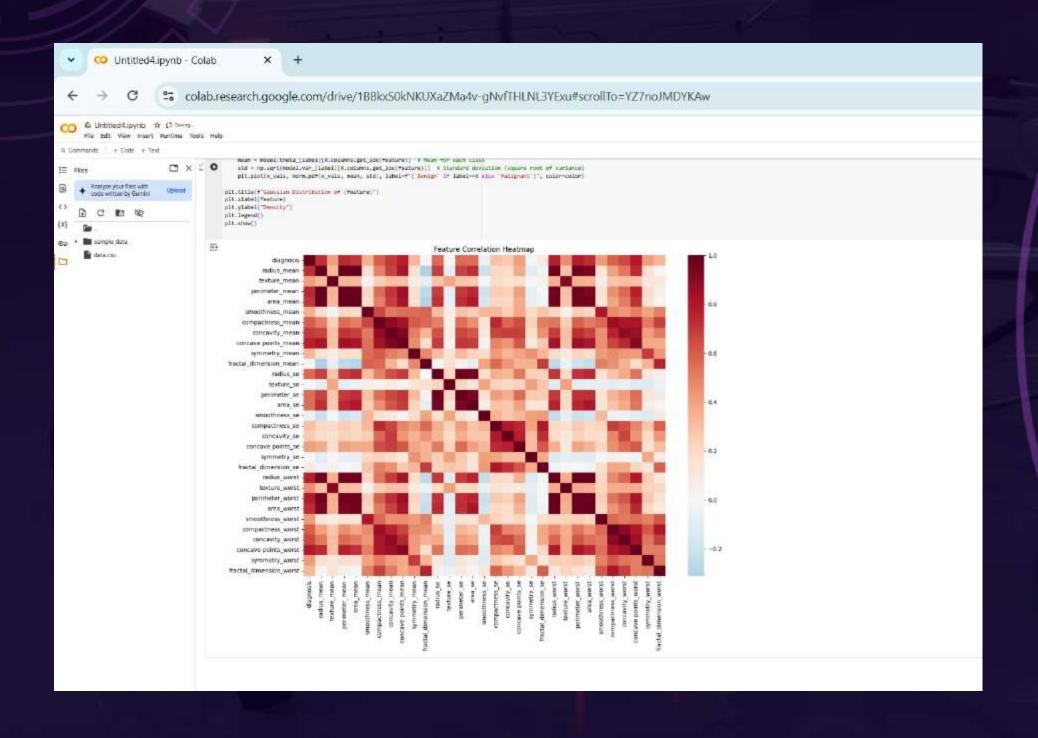














# Insights

- High Accuracy: Gaussian Naive Bayes achieved ~96% accuracy, showing strong performance in classifying tumors as benign or malignant.
- Fast & Efficient: The model trains and predicts quickly, making it suitable for real-time or resource-limited settings.
- False Negatives Are Low: The model showed high recall for malignant cases crucial in medical diagnosis.
- Performs Well Despite Assumptions: Even though it assumes feature independence and normal distribution, it still performs robustly

### Recommendation

Based on the findings, the following recommendations were made:

- Clinical Integration: Use the model as a decision-support tool in diagnostic workflows.
- Validate on Real-World Data: Test with diverse and noisy datasets to ensure reliability beyond the benchmark dataset.
- Confidence Thresholding: Flag low-confidence predictions for manual review to reduce risk.
- Enhance SEO efforts to maximize organic traffic acquisition

