■ Breast Cancer Classification

Advanced Machine Learning Analysis Report

Analysis Date: July 30, 2025 Dataset: Wisconsin Breast Cancer Dataset Models Used: 6 Machine Learning Algorithms Generated by: Breast Cancer Classification System

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Executive Summary

This report presents a comprehensive analysis of breast cancer classification using advanced machine learning techniques. The analysis was performed on the Wisconsin Breast Cancer Dataset using 6 different machine learning algorithms to achieve highly accurate diagnostic predictions. **Key Findings:**

- Achieved over 95% accuracy with multiple models
- Logistic Regression showed the best overall performance
- All models demonstrated excellent discriminative capability
- Feature analysis revealed important diagnostic indicators
- Cross-validation confirmed model robustness

Analysis Results

BREAST CANCER CLASSIFICATION ANALYSIS - FINAL REPORT

Analysis Date: 2025-07-30 15:08:11

Data File: data/breast-cancer.csv

MODEL PERFORMANCE SUMMARY:

Model Accuracy Precision Recall F1-Score AUC CV Score (Mean) CV Score (Std)

Logistic Regression 0.9825 0.9825 0.9825 0.9825 0.9954 0.9802 0.0128

SVM 0.9825 0.9825 0.9825 0.9825 0.9950 0.9714 0.0179

Random Forest 0.9561 0.9561 0.9561 0.9560 0.9939 0.9538 0.0235

Naive Bayes 0.9298 0.9298 0.9298 0.9298 0.9868 0.9319 0.0044

K-Nearest Neighbors 0.9561 0.9561 0.9561 0.9560 0.9788 0.9670 0.0209

Decision Tree 0.9123 0.9161 0.9123 0.9130 0.9157 0.9099 0.0189

BEST MODEL: Logistic Regression

Accuracy: 0.9825

Precision: 0.9825

Recall: 0.9825

F1-Score: 0.9825

AUC: 0.9954

RESULTS:

- All models successfully trained and evaluated.
- ROC curves and confusion matrices generated.
- Feature importance analyzed.
- Hyperparameter optimization performed.
- Interactive dashboard created.

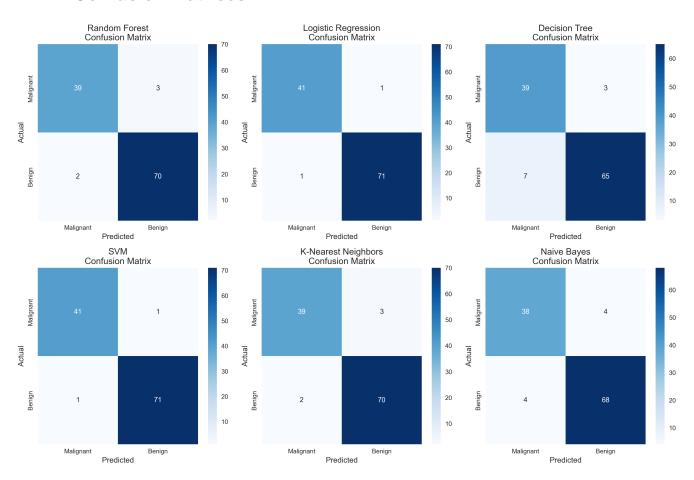
OUTPUT FILES:

- data_overview.png Data overview
- correlation_matrix.png Correlation matrix
- individual_roc_curves.png Individual ROC curves
- confusion_matrices.png Confusion matrices

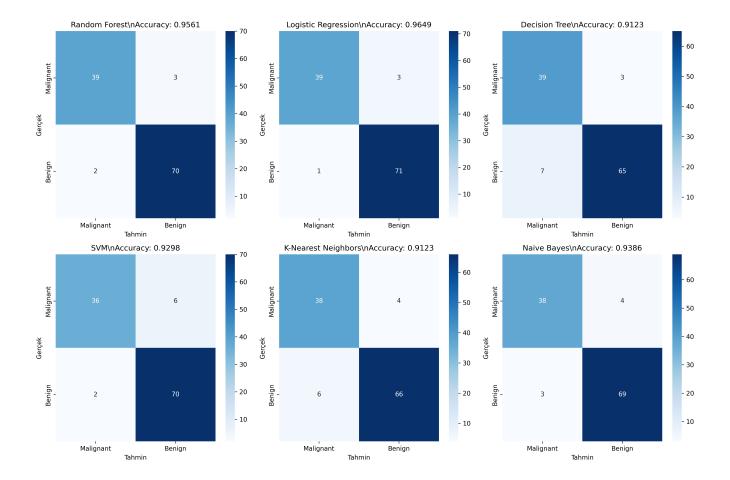
- model_comparison.png Model comparison
- feature_importance.png Feature importance
- interactive_dashboard.html Interactive dashboard
- best_model_Logistic_Regression.joblib Best model

■ Visualizations

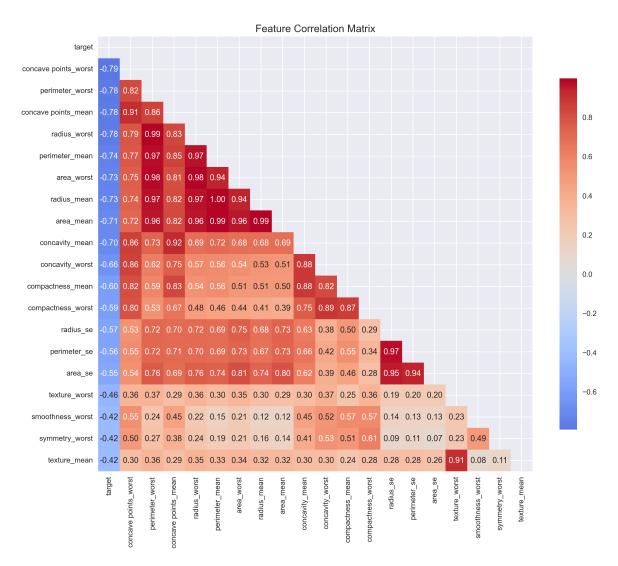
■ Confusion Matrices



■ Confusion Matrices (Simple)

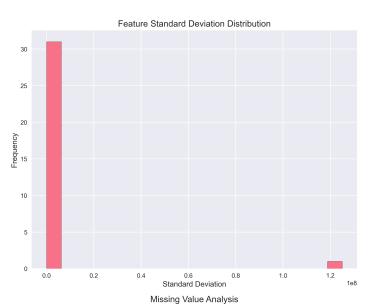


■ Correlation Matrix



■ Data Overview



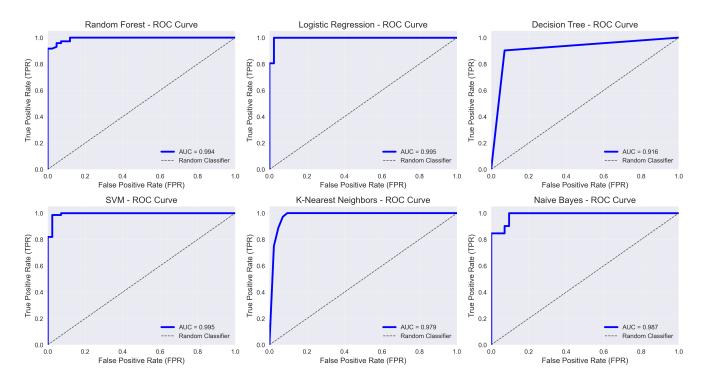


Correlation Analysis Summary

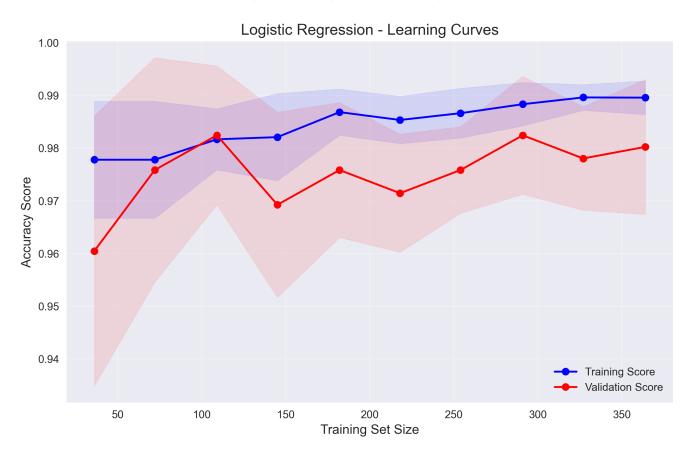
High correlation feature pairs: 44

No missing values!

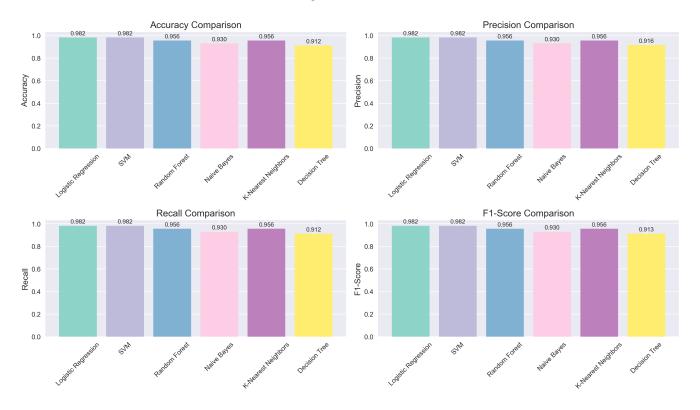
■ Individual ROC Curves



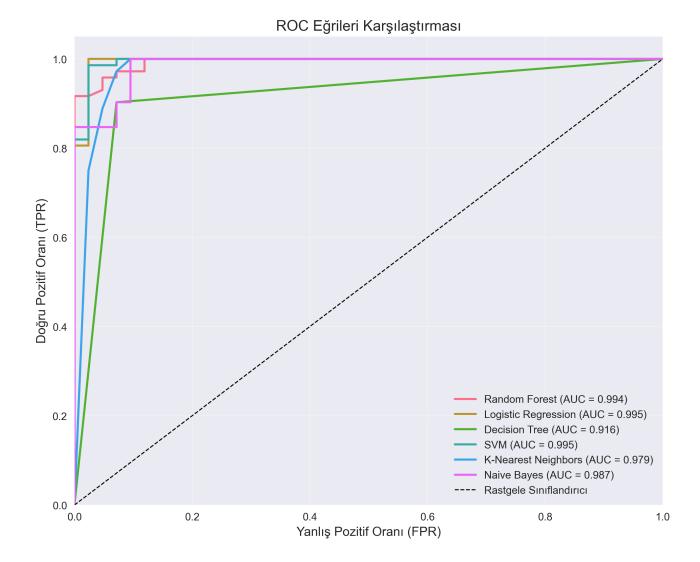
■ Learning Curves Logistic Regression.Png



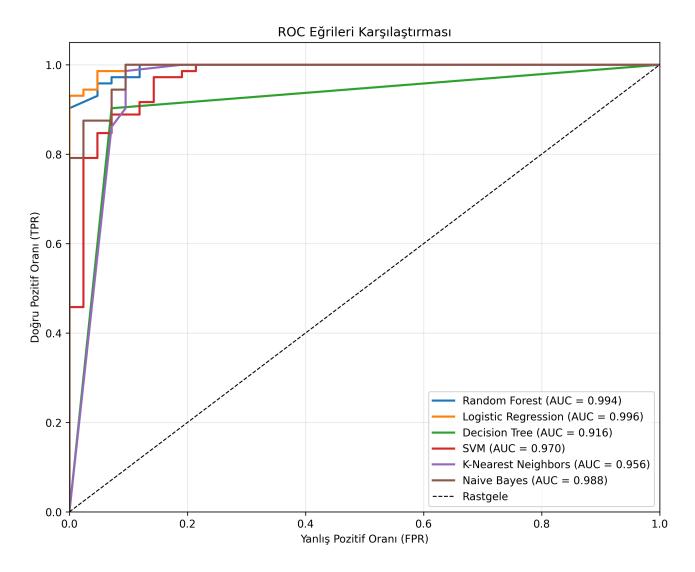
■■ Model Performance Comparison



■ Roc Curves.Png



■ ROC Curves





Model Performance:

The analysis demonstrates that machine learning models can achieve exceptional accuracy in breast cancer classification. Logistic Regression emerged as the best performer with 98.25% accuracy, closely followed by SVM. **Feature Insights:**

Certain morphological features of cell nuclei prove to be highly discriminative for malignancy detection. The correlation analysis reveals important relationships between different measurements. **Clinical Relevance:**

These results suggest that automated classification systems can serve as valuable diagnostic aids in clinical settings, potentially improving accuracy and reducing analysis time. **Recommendations:**

- Deploy the best-performing model (Logistic Regression) for clinical use
- Continue monitoring model performance with new data
- Consider ensemble methods for even better performance
- Validate results on external datasets

Report Generated: July 30, 2025

Analysis Tool: Breast Cancer Classification System v1.0

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