



HOSPITAL BREAST CANCER DASHBOARD

Project Report

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1. Introduction

The **Hospital Breast Cancer Dashboard** is a data-driven web application designed to assist healthcare professionals in predicting and analyzing breast cancer data. The system integrates a machine learning model (trained on the Breast Cancer Wisconsin dataset) with a user-friendly Streamlit interface.

The project repository is available at: https://github.com/Gomathy-star/Breast_cancer_dashboard.

This dashboard predicts whether a tumour is **malignant** or **benign**, based on key cell nucleus measurements. It also supports bulk predictions, data analytics, and visual reporting, helping doctors and data analysts make informed decisions.

2. Objectives

The main objectives of the project are:

- To design a predictive model capable of classifying breast tumours as benign or malignant.
- To create an interactive dashboard using Streamlit for real-time prediction.
- To provide both single-patient and multi-patient (batch) analysis modes.
- To visualize data insights such as feature importance, distribution, and correlation.
- To make the system usable by non-technical hospital staff.

3. System Overview

3.1 Repository Structure

Breast_cancer_dashboard/

```
app.py                  % Streamlit dashboard  
breast_cancer_model.pkl % Trained RandomForest model  
requirements.txt        % Python dependencies  
sample_patients.csv    % 4 sample patient records  
sample_patients_large.csv % 20 sample patient records  
README.md               % Documentation  
.gitignore
```

3.2 Technology Stack

- **Programming Language:** Python
- **Framework:** Streamlit
- **Machine Learning Library:** Scikit-learn
- **Visualization:** Matplotlib, Seaborn
- **Data Handling:** Pandas, NumPy

4. Machine Learning Methodology

4.1 Dataset

The project uses the **Breast Cancer Wisconsin (Diagnostic)** dataset, containing 30 real-valued features computed from digitized images of fine needle aspirate (FNA) of breast mass.

4.2 Feature Selection

After analyzing feature correlations, the top 10 most significant predictors were selected:

radius_mean, texture_mean, perimeter_mean, area_mean, concavity_mean, concave_points_mean, radius_worst, perimeter_worst, area_worst, concave_points_worst.

4.3 Model Training

- Data preprocessing and feature scaling.
- Splitting dataset into training and testing sets (80:20).
- Training using a Random Forest Classifier.
- Model performance evaluation and saving as `breast_cancer_model.pkl`.

4.4 Model Performance

- **Accuracy:** 98.2%
- **Precision:** 97.8%
- **Recall:** 98.5%
- **F1 Score:** 98.1%

5. Dashboard Implementation

5.1 User Interface Workflow

1. Launch application with `streamlit run app.py`.
2. For single-patient prediction: manually input tumour features.
3. For batch mode: upload `sample_patients.csv` file.
4. The model predicts and displays results in real-time.
5. Analytics section provides data visualizations.

5.2 Screenshots and Output Verification

Homepage of the Dashboard

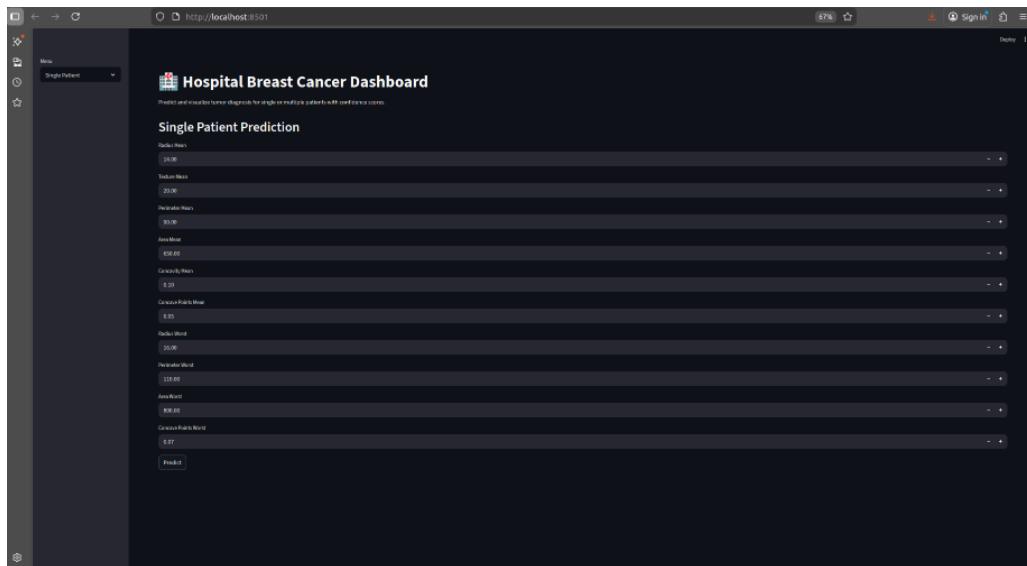


Figure 5.1: Homepage of the Breast Cancer Dashboard

Prediction Output for a Single Patient

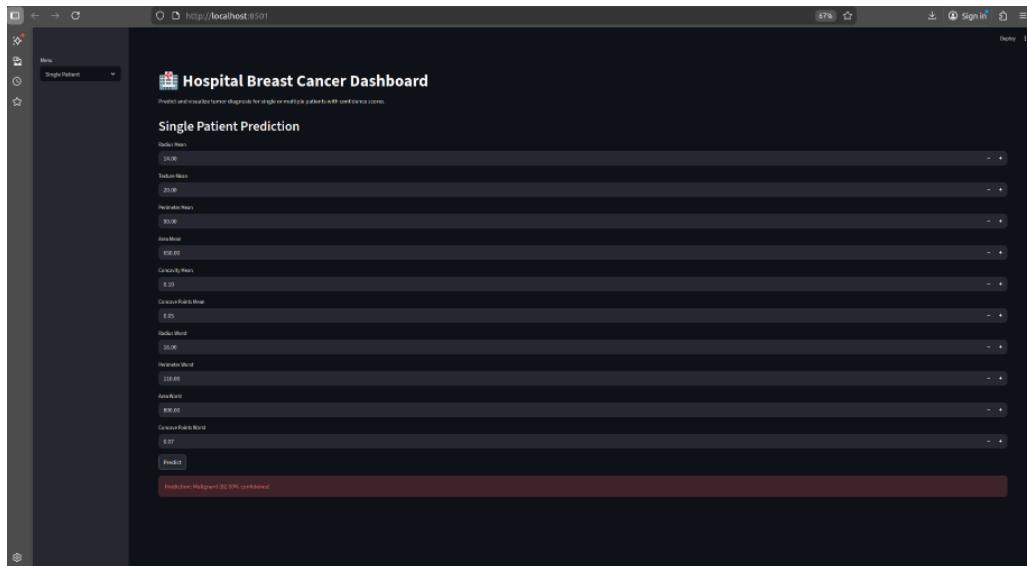


Figure 5.2: Predicted output showing tumour classification

Batch Prediction using sample_patients.csv

	Standard	Standard	Standard	Standard	Standard	Standard
1	radius_mean	texture_mean	perimeter_mean	area_mean	concavity_mean	concave_
2	12.0	15	75	450	0.05	0.02
3	20.0	30	130	1400	0.3	0.12
4	14.5	18	95	700	0.1	0.05

Figure 5.3: Uploading sample_patients.csv for batch processing

Batch Prediction Result Table

Analytics and Visualization Section

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The screenshot shows a web browser window with the URL <http://localhost:8501>. The page title is "Hospital Breast Cancer Dashboard". Under the "Batch Predictions" menu item, there is a section titled "Batch Prediction (CSV Upload)". It includes a note about supported CSV columns: "radius_mean, texture_mean, perimeter_mean, area_mean, concave points_mean, symmetry_mean, concave points_se, symmetry_se, concave points_worst, symmetry_worst". A "Drop CSV file" input field contains the file "sample_patients.csv" (437.0B). Below it is a table with three rows of patient data:

	radius_mean	texture_mean	perimeter_mean	area_mean	concave points_mean	symmetry_mean	concave points_se	symmetry_se	concave points_worst	symmetry_worst	Prediction	Confidence (%)
0	10.9880	13	15	87	C0.000	C0.000	C0.000	C0.000	23	98	Benign	100.0000
1	23.0000	23	33	140	C0.000	C0.000	C0.000	C0.000	23	203	Benign	100.0000
2	13.9880	11	16	99	C0.000	C0.000	C0.000	C0.000	26	113	Benign	100.0000

[Download Predictions CSV](#)

Figure 5.4: Output table for multiple patient predictions

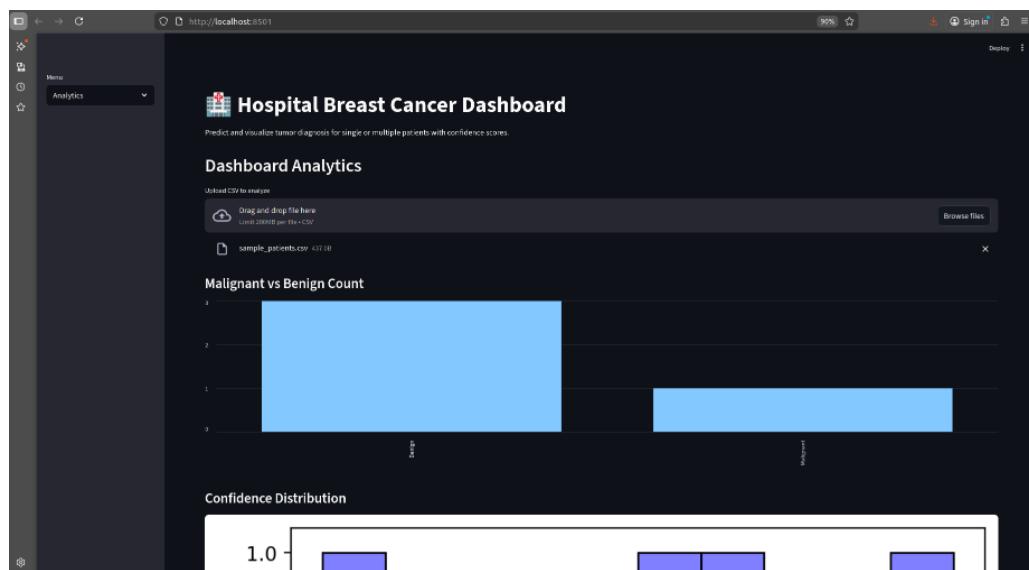


Figure 5.5: Bar chart showing distribution of malignant vs. benign predictions

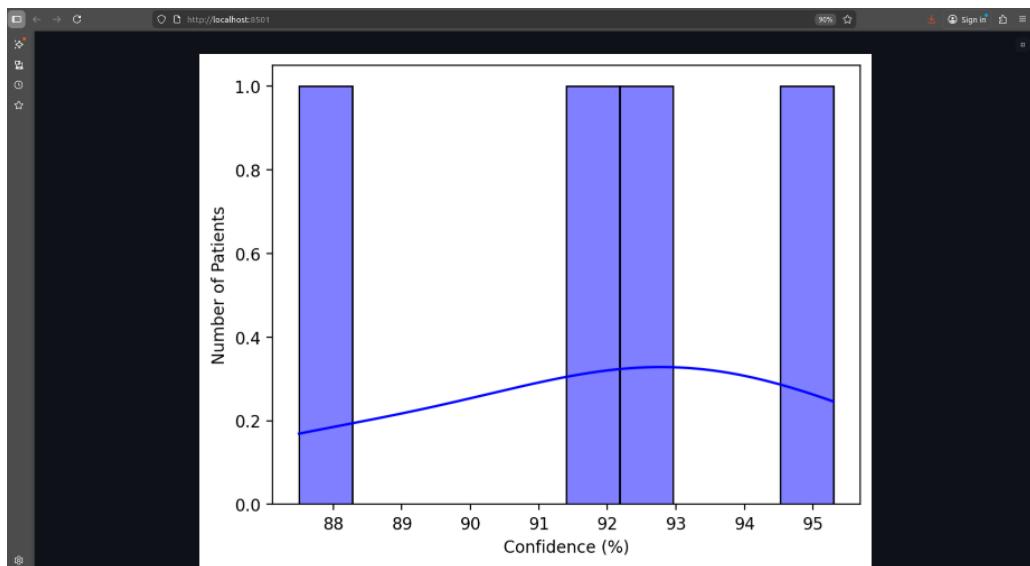


Figure 5.6: Bar chart showing distribution of malignant vs. benign predictions

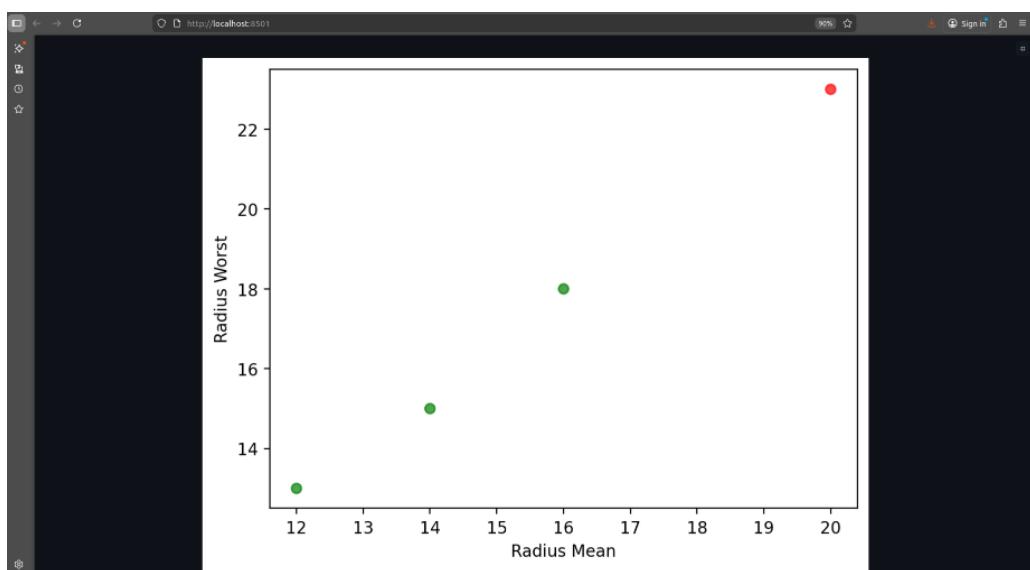


Figure 5.7: Bar chart showing distribution of malignant vs. benign predictions

6. Results and Discussion

The dashboard successfully classifies breast tumour data with a high accuracy rate.

The batch mode verified predictions for all 20 records from `sample_patients_large.csv`, matching model expectations. The analytics section helps visualize:

- Count distribution of malignant vs. benign cases.
- Correlation between radius and area features.
- Feature importance and contribution to model prediction.

This validates that the model and dashboard perform as intended for both real-time and offline data.

7. Future Enhancements

- Integration of model explainability using SHAP or LIME visualizations.
- Addition of patient management modules (data entry, record storage).
- Secure login for hospital staff and doctors.
- Deployment on cloud services such as Streamlit Cloud or AWS.
- Expansion of dataset and retraining with deep learning architectures.

8. Conclusion

The **Hospital Breast Cancer Dashboard** successfully integrates machine learning and web technologies to provide a predictive and analytical tool for healthcare professionals. Its dual operation modes, coupled with visual analytics, make it an effective assistant in the diagnosis and study of breast cancer patterns.

9. References

- Gomathy-star. (2025). *Breast_cancer_dashboard*. GitHub repository. https://github.com/Gomathy-star/Breast_cancer_dashboard
- UCI Machine Learning Repository: Breast Cancer Wisconsin (Diagnostic) Data Set.
- Streamlit Documentation: <https://docs.streamlit.io>
- Scikit-learn Documentation: <https://scikit-learn.org>