## # Al-Powered Bug Predictor — Final Report

## ## Project Summary

This project applies machine learning techniques to simulate a bug prediction system. Due to the unavailability of the 'pc1.csv' defect dataset, we use the breast cancer classification dataset to represent a binary classification problem and explore a structured AI development workflow.

# ## Objectives

- Build a supervised ML model using a known dataset.
- Analyze the feature importance in classification.
- Evaluate model accuracy and visualize insights.

# ## Methodology

- 1. \*\*Data Preparation\*\*:
  - Used 'sklearn.datasets.load breast cancer'.
  - Extracted feature matrix and labels.
  - Performed train-test split (80:20 ratio).
- 2. \*\*Model Training\*\*:
  - Used `RandomForestClassifier` for robustness and feature importance capability.
  - Trained model and calculated accuracy.
- 3. \*\*Feature Importance Analysis\*\*:

- Extracted top 10 most important features.
- Visualized using a horizontal bar chart.

### 4. \*\*Output\*\*:

- Achieved over 95% accuracy (sample output).
- Displayed feature importance of predictors like `mean concave points`, `worst concave points`, etc.

#### ## Results

The visualization shows features most influential in predicting outcomes. This is useful in software defect analysis, where understanding the root cause is as important as prediction.

![Feature Importance Output](./0ce83521-5229-43e5-ab87-c2792360b156.png)

### ## Challenges

- The 'pc1.csv' dataset could not be accessed (404 error).
- Used a substitute dataset for demonstration purposes.
- Visualization and logic still follow best practices for bug prediction systems.

#### ## Recommendations

- Replace mock data with a real defect dataset (e.g., NASA PROMISE).
- Incorporate explainability tools (e.g., SHAP).
- Add automation and deployment capability.

## ## Conclusion

The project successfully simulates a bug prediction pipeline. It demonstrates how ML models can analyze structured data to make accurate predictions and explain why specific features are important.

## ## Tools Used

- Google Colab
- Scikit-learn
- Pandas & Matplotlib
- Breast Cancer Dataset

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