

# **Report on K-Fold Cross-Validation in Machine Learning (Lab-5)**

## **K-Fold Cross-Validation in Machine Learning**

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### **1. Experiment Overview**

The lab focuses on the application of K-Fold Cross-Validation to evaluate the accuracy of various machine learning algorithms. The objective is to test the performance of five classification algorithms on three datasets:

- Pima Indians Diabetes Dataset
- Wine Quality Dataset
- Breast Cancer Wisconsin Dataset

### **2. Introduction**

In machine learning, model evaluation is critical to assess reliability. K-Fold Cross-Validation is a method that helps mitigate overfitting by dividing the dataset into K equal parts (folds). The model is trained on K-1 folds and tested on the remaining fold, providing a comprehensive measure of performance.

The algorithms used are:

- Logistic Regression
- Decision Tree
- Support Vector Machine (SVM)
- K-Nearest Neighbors (KNN)
- Linear Discriminant Analysis (LDA)

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### **3. Datasets Used**

- Pima Indians Diabetes Dataset: Medical data of women for predicting diabetes.
- Wine Quality Dataset: Data related to wine characteristics for quality classification.
- Breast Cancer Wisconsin Dataset: Features related to cancer cells for malignancy prediction.

### **4. Steps in Experiment**

1. Importing Packages: Necessary libraries (Pandas, Matplotlib, Sklearn) were used for analysis.
2. Loading the Dataset: Each dataset was loaded and verified using Pandas.
3. Data Splitting: Features and target variables were separated and split for training/testing.
4. Model Definition: Logistic Regression, LDA, KNN, Decision Tree, Naive Bayes, and SVM were initialized.
5. Cross-Validation: Stratified K-Fold Cross-Validation (10 folds) was used to evaluate models.
6. Visualization: A boxplot was used to compare model accuracy scores.

### **5. Results**

The Support Vector Classifier (SVC) achieved the highest accuracy across the folds, suggesting its effectiveness in predicting outcomes.

### **6. Practical Application**

A new patient's data was tested with the SVC model, successfully predicting the likelihood of diabetes. This highlights the potential of machine learning in healthcare.

### **7. Conclusion**

K-Fold Cross-Validation is a robust method for evaluating models. SVC performed best among the algorithms tested, particularly for medical predictions.