THE ROLE OF MACHINE LEARNING IN PREDICTING DIABETES

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This research evaluates the effectiveness of machine learning models—Logistic Regression, Decision Tree, and XGBoost—in predicting diabetes. By analyzing patient data, including glucose, BMI, and age, the study identifies key risk factors and highlights how ML can improve early detection and intervention.

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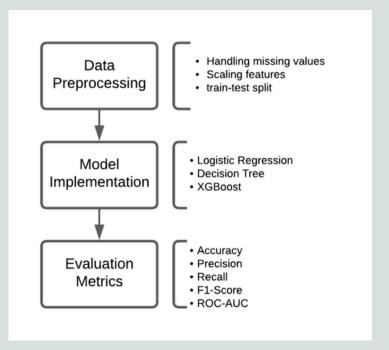
INTRODUCTION

This study explores how machine learning can help predict diabetes early and identify key factors like glucose levels, BMI, and age. By comparing three models—Logistic Regression, Decision Tree, and XGBoost—this research shows that XGBoost is the most accurate, achieving a 98% success rate. The findings highlight the potential of machine learning to support healthcare professionals in early detection and personalized interventions, offering a practical solution to tackle one of the world's most prevalent chronic diseases.

METHODOLOGY

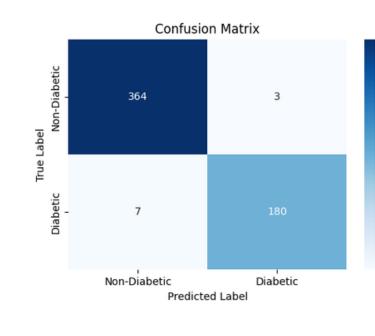
Three models were evaluated using key performance metrics on a dataset of $\sim 3,000$ records. Preprocessing included imputing missing values and standardizing features.

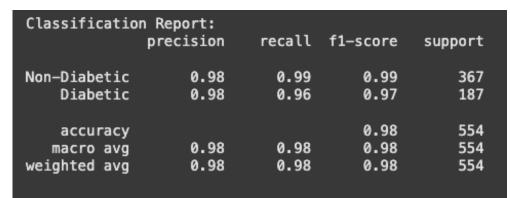


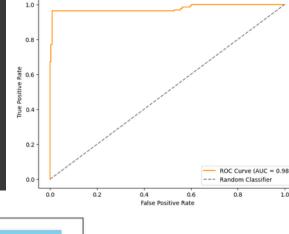


RESULTS

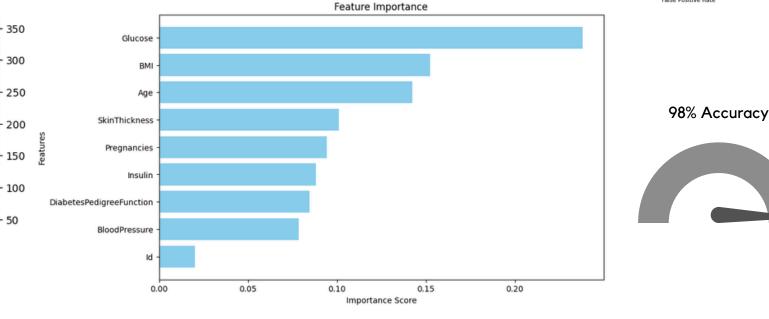
XGBoost achieved the highest accuracy (**98.19**%) and demonstrated superior performance in all metrics. Glucose, BMI, and Age were identified as the top predictors of diabetes.







Receiver Operating Characteristic (ROC) Curve



FINDINGS

- Glucose levels and BMI are critical, modifiable predictors of diabetes.
- Dietary changes like ketogenic diets and regular exercise can mitigate risks.
- XGBoost is a reliable tool for predictive healthcare applications.
- Future work should focus on more diverse datasets and explainable AI techniques.

CONCLUSION

- Glucose, BMI, and age are key predictors, highlighting the importance of lifestyle interventions like low-carb diets and exercise.
- XGBoost proved to be the most reliable model for accurate diabetes predictions.
- Future work should focus on diverse datasets and explainable AI for better clinical adoption.
- These findings can guide healthcare professionals in early detection and personalized treatment strategies.

RELATED LITERATURE/REFERENCE

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