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**Enhancing proactive patient outreach with medical message classification using synthetic data from ChatGPT**

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**PROIBIDO A REPRODUÇÃO, APENAS PARA CONSULTA.**

# PAPER 34 ENHANCING PROACTIVE PATIENT OUTREACH WITH MEDICAL MESSAGE CLASSIFICATION USING SYNTHETIC DATA FROM CHATGPT

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## INTRODUCTION

The growing need for proactive patient outreach has driven the use of automated message classification. While advanced tools exist, their high costs limit small companies. Traditional, cost-effective methods remain a viable alternative.

## OBJECTIVE

This article aims to identify the effect of data augmentation with synthetic data generated by ChatGPT 3.5 on the performance of traditional and Transformer models for the binary classification of health insurance user messages regarding the presence or absence of medical needs.

## METHODOLOGY

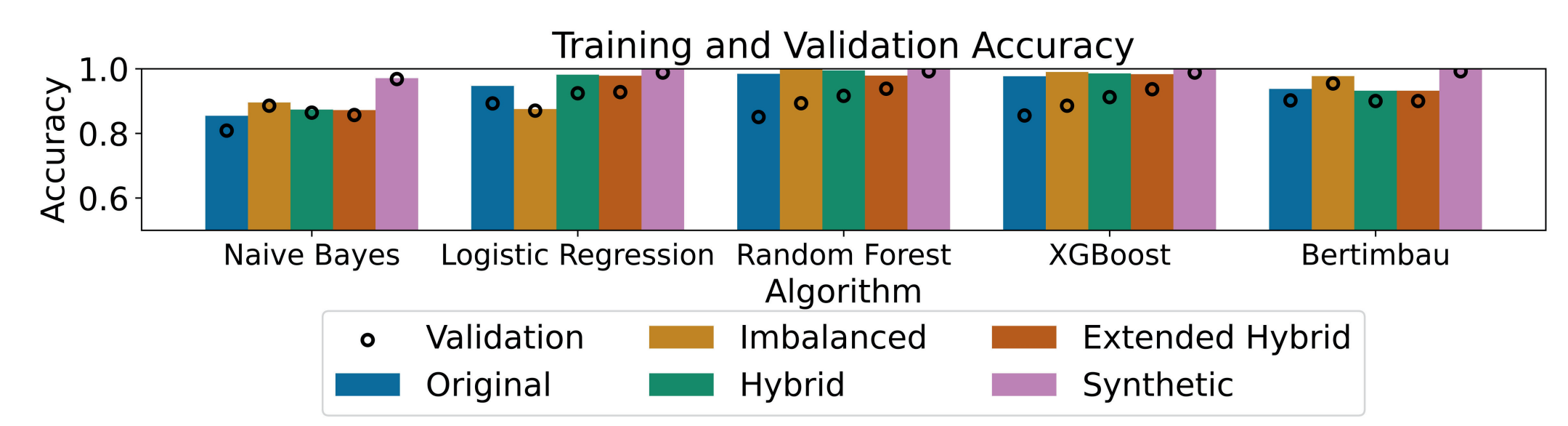
The primary aim is to classify messages that indicate medical needs, aiding in patient outreach and efficient resource allocation. We applied Natural Language Processing (NLP) techniques, including Logistic Regression, Naive Bayes, Random Forest, XGBoost, and BERTimbau, a Portuguese transformer-based model. Models were trained on both original and augmented datasets.



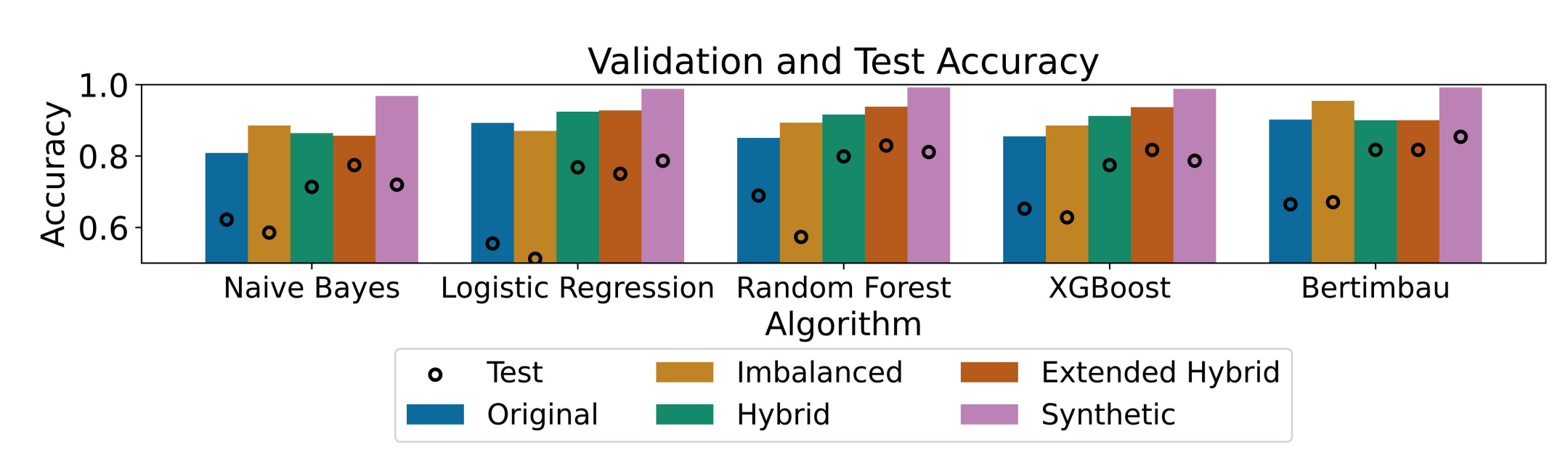
## DATASET

A. Camargo de Brito, A. Galindo Leal, G. Torres Custódio, E. José Rodrigues, M. Marcia Viana Martins, and V. Monteiro de Paula Guirado, "Dataset for the IEEE CIHM2025 paper "Enhancing Proactive Patient Outreach with Medical Message Classification Using Synthetic Data from ChatGPT", Dec. 2024. [Online]. Available: <https://github.com/IPT-TD-SIAA/IEEECIHM2025.git>

### 01 Training and Validation Accuracy

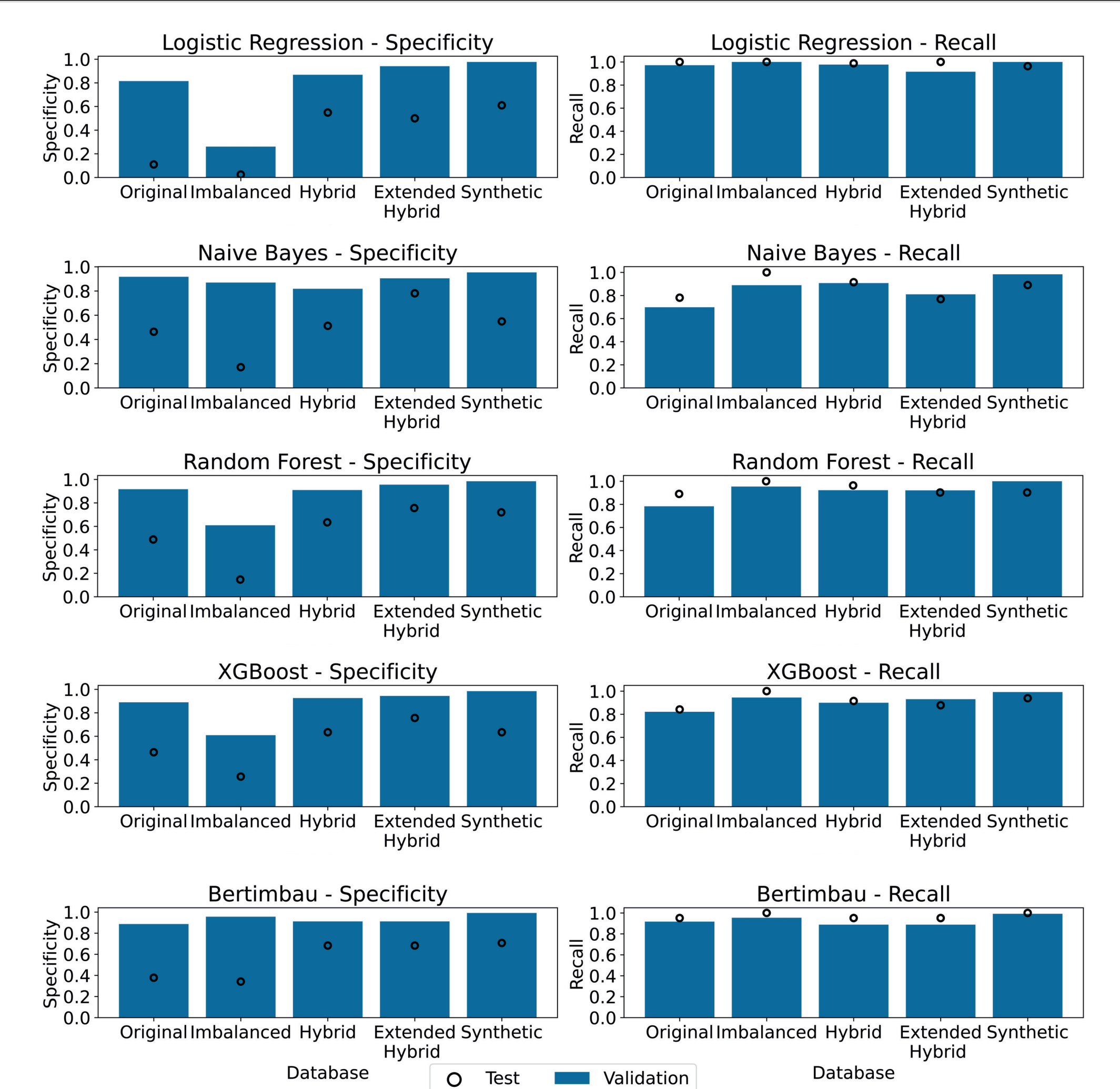


### 02 Validation and Test Accuracy



### 03 Specificity and Recall

In binary classification, messages with medical needs are labeled as "one" and without as "zero." Models trained on original or imbalanced datasets struggled to correctly identify messages without medical needs, especially with simpler algorithms. Hybrid and synthetic datasets improved adaptability, though capturing subtle nuances in patient communication remained challenging.



## ANALYSIS

Models trained on all datasets exhibited similar training and validation accuracies, with some overfitting observed in traditional approaches. The use of synthetic data helped mitigate accuracy gaps, enhancing model generalization. Hybrid and synthetic datasets resulted in the highest validation accuracy, particularly in models addressing imbalanced data. Models trained on hybrid and synthetic datasets demonstrated superior generalization, maintaining higher accuracy on unseen data. Specificity and recall varied across models, with synthetic data contributing to improved classification performance. Overall, hybrid and synthetic datasets enhanced model robustness.

## CONCLUSION

This study demonstrated that GPT-generated synthetic data improves medical message classification by enhancing model accuracy, reducing overfitting, and improving generalization. The BERTimbau model showed significant performance gains with synthetic data, making it a valuable strategy for addressing class imbalance. These findings highlight the potential of synthetic data to optimize automated medical message processing and enhance operational efficiency. Future research should explore its application across different languages and medical subfields to further validate its effectiveness.

