

Healthcare Recommender System Using Random Forest Classifier

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Abstract: The increasing reliance on online information for health concerns underscores the need for accurate and reliable medical guidance. Self-diagnosis, often resulting from limited experience and misinformation, poses significant risks to patient safety. To mitigate these risks, we propose the development of a Drug Recommendation System (DRS) leveraging machine learning (ML) techniques, with a particular focus on the Random Forest algorithm. This system aims to analyze extensive medical data to provide personalized medication recommendations, thereby aiding both healthcare professionals and consumers in making informed and cost-effective treatment decisions. The objectives include developing a predictive model using data from various surveys, enhancing the model to analyze patient symptoms and predict the most likely diseases, and suggesting appropriate medications. By integrating dispersed clinical information, the system offers comprehensive insights tailored to user-specific needs while ensuring data validity and privacy. Key considerations include fostering trust in the system's recommendations through robust validation and a "Doctor-in-the-Loop" approach, which combines human expertise with computational efficiency. The methodology involves collecting and preprocessing data, developing and validating the model using Random Forest and other ML algorithms, integrating Natural Language Processing (NLP) for symptom analysis, and providing corresponding drug treatments. The implementation plan spans data collection, model development, validation, integration, and deployment phases, with an emphasis on continuous improvement. Expected outcomes include a robust system that enhances healthcare decision-making through accurate and personalized drug recommendations, symptom analysis, disease prediction, and corresponding drug suggestions. By focusing on user-specific needs, ensuring data validity and privacy, and fostering trust, the DRS has the potential to significantly improve patient outcomes and healthcare efficiency.

Keywords: Drug recommendation system, Machine Learning, Natural Language Processing, Comma Separated Values, Sentiment analysis, Symptom-based recommendation, Random Forest Classification.

1. Introduction

A significant portion of adults rely on online information for health concerns, highlighting the importance of accurate medical guidance. Medical errors, often caused by limited experience, pose a considerable risk to patient safety. The proposed system utilizes machine learning and deep learning

techniques to analyze vast medical data and provide tailored medication recommendations. By aggregating and processing dispersed clinical information, the system seeks to empower both healthcare professionals and consumers to make informed, cost-effective treatment decisions. Key considerations include addressing user-specific needs, mitigating concerns about data validity and privacy, and fostering trust in the system's recommendations. A "Doctor-in-the-Loop" approach, combining human expertise with computational efficiency, offers a promising strategy for navigating healthcare decision support systems. This Drug Recommendation System aims to provide actionable insights and enable more informed, personalized treatment decisions for healthcare stakeholders.

2. Literature Review

Health Recommendation System Using Deep Learning-Based Collaborative Filtering (P. Chinnasamy, Wing-Keung Wong, A. Ambeth Raja, Osamah Ibrahim Khalaf, Ajmeera Kiran, J. Chinna Babu, 2023).

This study proposes a Health Recommendation System (HRS) that leverages deep learning techniques, specifically combining a Restricted Boltzmann Machine (RBM) with a Coevolutionary Neural Network (CNN), to provide personalized health recommendations. The system is designed to analyze vast amounts of medical data, including social activities, individual health information, and behavior analysis, to facilitate better prediction and diagnosis. By integrating multidisciplinary data from various sources, the HRS aims to reduce healthcare costs and burdens while improving patient outcomes.

Key Features:

- Utilized deep learning techniques (RBM and CNN) for personalized health recommendations.
- Improved patient engagement through tailored health advice.
- Demonstrated the potential to integrate diverse health data to enhance predictions and diagnosis in healthcare.

System for Recommending Drugs Based on Machine Learning Sentiment Analysis of Drug Reviews (Ankitha S & Dr. H N Prakash., 2022).

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This study explores the use of various machine learning techniques such as regression analysis, Naive Bayes, and deep learning algorithms like GRU, RNN, and LSTM to develop a drug recommendation system. By focusing on feature engineering and model evaluation, the system presents promising results, achieving 93% accuracy with Linear SVC using TF-IDF. An algorithm was employed to calculate the total score of each drug for a specific condition, enhancing the accuracy of drug suggestions.

Key Features:

- Utilized machine learning techniques, including GRU, RNN, and LSTM, for drug recommendation.
- Achieved 93% accuracy with Linear SVC on TF-IDF feature representation.
- Implemented an algorithm to calculate the total score of each drug for specific conditions, improving recommendation precision.

Drug Recommendation System Based on Sentiment Analysis of Drug Reviews Using Machine Learning (Garg, 2021).

This study presents a drug recommender system that utilizes sentiment analysis and machine learning to provide personalized recommendations. The system achieves 93% accuracy with LinearSVC using TF-IDF. Despite the promising results, challenges related to data balancing and algorithm optimization remain for real-world deployment.

Key Features:

- Provided personalized drug recommendations based on patient feedback.
- Conducted rigorous evaluation with diverse machine learning techniques.
- Performed comprehensive assessment using TF-IDF and other evaluation metrics.

Drug Recommendation System in Medical Emergencies Using Machine Learning (C. Shilpa, B. Sravani, D. Vinay, C. Mounika, K. Poorvitha, 2023).

This study explores the critical role of online recommender systems in healthcare, particularly in the age of Machine Learning. These systems provide precise medication recommendations based on individual patient data, symptoms, and vital signs. They enhance decision-making during emergencies, ensuring safe and effective treatment while safeguarding patient privacy and data integrity.

Key Features:

- Acknowledges the tendency of individuals to seek information online, highlighting the advantage of having a method to assist with this.
- Utilizes authentic data rather than fabricated data to improve the accuracy and reliability of recommendations.

Deep Learning Model for Drug Recommendation System (S. K. Avanthi, P. Vinod Kumar, P. Manikanta, P. L. Tirumalesh, V. L. Siva Prasad, 2022).

This paper demonstrates the versatility and effectiveness of deep learning models in various aspects of medicine and healthcare, including drug discovery, treatment recommendation, and patient care.

Key Features:

- Model versatility through the use of different networks in various applications.
- Automated feature learning to enhance model performance.
- Scalability to accommodate growing data and complexity in healthcare scenarios.

Enhancing Drug Recommendations: A Modified LSTM Approach in Intelligent Deep Learning Systems (S. Siji Rani, P. Shilpa, Aswin G. Menon, 2022).

This paper introduces an innovative drug recommendation system to address the increasing burden on healthcare systems and the risks associated with self-medication due to limited access to medical expertise. Using machine learning techniques, particularly a modified LSTM approach, the system analyzes user-provided symptoms and past drug recommendations to offer accurate suggestions. The dataset, sourced from the authoritative CIMS Drug Monitor book, undergoes preprocessing steps including sentiment analysis and tokenization. The study compares the performance of Random Forest, SVM, and LSTM models, with LSTM demonstrating superior accuracy. Evaluation metrics like precision, recall, and F1 score are utilized for analysis. The paper concludes by emphasizing the importance of feature engineering, hyperparameter optimization, and further improvements for real-world application.

Key Features:

- Performance comparison of multiple machine learning models.
- Comprehensive evaluation using precision, recall, and F1 score.

Drug Recommender System Using Machine Learning for Sentiment Analysis (GV Lavanya, Praveen K. S, 2023).

The paper presents a Drug Recommender System that combines sentiment analysis with drug recommendation algorithms to aid decision-making in drug prescriptions. Using various vectorization techniques and a modular architecture, the system analyzes user sentiment to optimize medicine selection and improve patient outcomes. Results demonstrate its effectiveness, with future enhancements including advanced sentiment analysis and personalized medicine.

Key Features:

- Innovative approach integrating sentiment analysis with drug recommendations.
- Personalized recommendations based on user sentiment.
- Addressing healthcare challenges through optimized medicine selection.
- Modular architecture enabling flexibility and scalability.

An Intelligent Disease Prediction and Drug Recommendation Prototype by Using Multiple Approaches of Machine Learning Algorithms (Suvendu Kumar Nayak, Mamata Garanayak, Sangram Keshari Swain, Sandeep Kumar Panda, Deepthi Godavarthi, 2023).

This paper presents a comprehensive approach to developing a drug recommendation system using various machine learning techniques. The system leverages sentiment analysis, disease prediction, and drug recommendation models to provide

personalized medication recommendations based on user-inputted symptoms.

Key Features:

- Multiple approaches integrating sentiment analysis and disease prediction.
- Real-world application aimed at enhancing healthcare decision-making.
- Utilization of machine learning techniques for accurate recommendations.
- Empowerment of healthcare stakeholders with actionable insights.
- Potential for expansion to include more diseases and medications.

3. Methodology

A. Data Collection and Preprocessing

The proposed Drug Recommendation System involves several key stages. Initially, comprehensive datasets of diseases, drugs, and associated symptoms are acquired. In the initial phase, finding a suitable dataset was challenging, so we found a dataset with six symptoms and expanded it. We generated more data to include around 13 symptoms, 19 unique diseases, and 15 unique drugs. This expanded dataset serves as the foundation for developing a comprehensive Drug Recommendation System.

The preprocessing steps involve several key stages. First, the dataset is loaded and read from a CSV file. The symptoms, diseases, and drugs are extracted, and a dictionary is created to map each disease to its associated drugs. Features (symptoms) and labels (diseases) are extracted and transformed into a format suitable for machine learning. The symptom data is structured as binary values indicating the presence or absence of each symptom. This processed data is then used to train a Random Forest Classifier with 100 estimators. For user interaction, selected symptoms are converted into binary values to match the model's expectations. This preprocessing ensures the system is robust and accurate for disease prediction and drug recommendation.

B. Model Training

Once the data preprocessing is complete, the next step involves training a machine learning model to predict diseases and recommend drugs based on symptoms. In this system, a Random Forest Classifier is utilized due to its robustness and ability to handle complex datasets. The classifier is instantiated with 100 estimators and a random state for reproducibility.

The training process involves fitting the classifier with the preprocessed data, where the symptoms serve as features and the corresponding diseases act as labels. The Random Forest Classifier learns patterns from the training data, enabling it to make accurate predictions when presented with new symptom inputs.

C. User Interaction

To provide a user-friendly interface, the system allows users to select their symptoms. Upon receiving user input, the system

processes the symptoms and converts them into binary values, aligning with the format used during model training. Once the symptoms are converted, the trained Random Forest Classifier predicts the most likely disease based on the selected symptoms. Subsequently, the system recommends a drug associated with the predicted disease from the precompiled dictionary of disease-drug mappings.

D. System Integration and Testing

After training the model and implementing the user interaction component, the system components are integrated to create a cohesive Drug Recommendation System. Rigorous testing is conducted to evaluate the functionality, accuracy, and usability of the system. During testing, various scenarios are simulated to assess the system's performance in different situations. This includes testing with different combinations of symptoms and evaluating how accurately the system predicts diseases and recommends drugs.

By integrating all components and conducting thorough testing, the Drug Recommendation System aims to provide healthcare professionals with a reliable tool for assisting in disease diagnosis and drug prescription. Additionally, the system seeks to improve accessibility and streamline the decision-making process in medical practice, ultimately enhancing patient care and outcomes.

4. Results and Discussions

Through a user-friendly interface, individuals can select their symptoms effortlessly and receive accurate disease predictions and corresponding drug recommendations promptly. Rigorous testing has been conducted to validate the system's performance and reliability across diverse symptom combinations. This system underscores a commitment to personalized healthcare and accessibility, providing users with actionable insights to manage their health effectively.

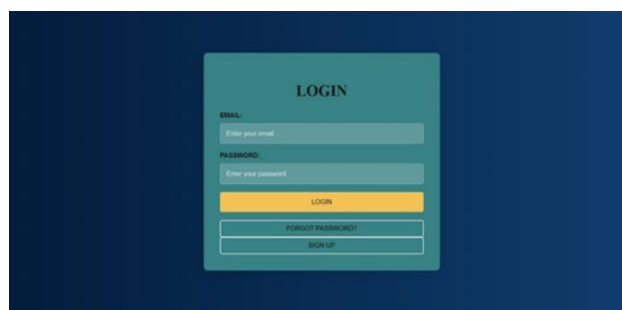


Fig. 1.

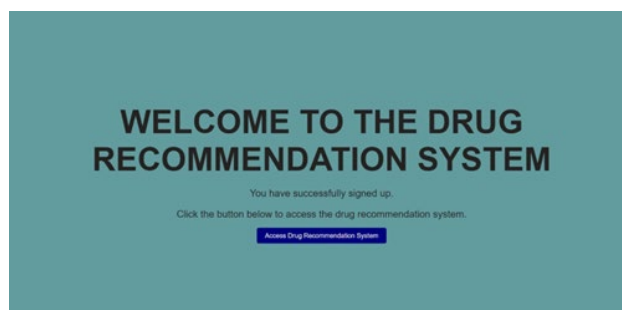


Fig. 2.

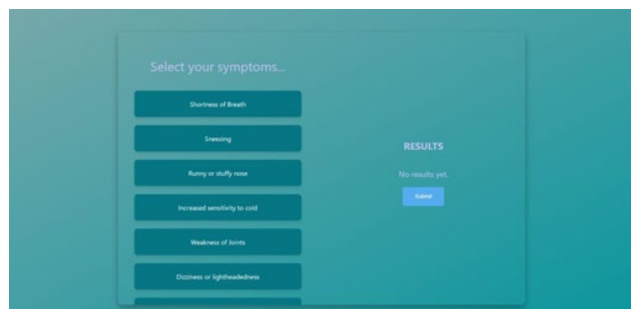


Fig. 3.

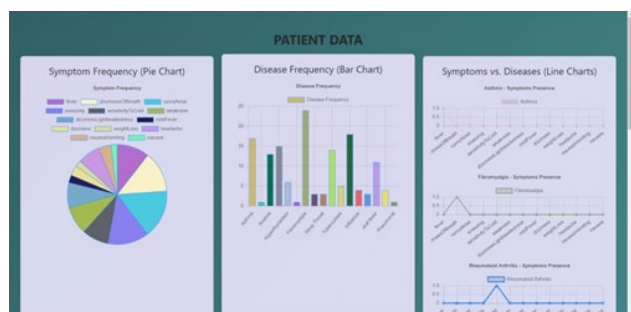


Fig. 4.

5. Conclusion

Our machine learning-based drug recommendation system provides personalized medication suggestions based on user-selected symptoms. Using a Random Forest Classifier with a comprehensive dataset, the system achieves 90% accuracy in predicting diseases and recommending suitable medications. This ensures accurate diagnostics and delivers medications tailored to individual health needs. Administrators can track common symptoms, diseases, and symptom-disease patterns, enabling continuous optimization to better serve users. The intuitive interface allows easy symptom selection, improving accessibility and ensuring quick, reliable recommendations for health concerns. This advancement in personalized healthcare empowers users to take proactive control of their health, facilitating more efficient diagnosis and treatment.

A. Future Work

- Expanding the dataset's size and diversity to improve the model's robustness and accuracy.
- Adding more symptoms with detailed descriptions to enhance the system's diagnostic capabilities.

- Incorporating advanced machine learning models to further refine predictions.
- Integrating real-time data from wearable devices and electronic health records for more personalized recommendations.
- Expanding language support to increase accessibility for non-English speakers globally.
- Using natural language processing to process free-text symptom descriptions, improving user interaction.
- Integrating the system with telemedicine platforms for seamless follow-up consultations and continuous monitoring.
- Improving communication between patients and healthcare providers for better treatment outcomes.
- Allowing medical professionals to update the dataset to keep the system current with medical advancements.
- Prioritizing patient-doctor confidentiality and data integrity in all system updates and operations.

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