Synopsis on

Medicine Recommendation System

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Title of Project:

Performing Sentiment Analysis on Medicine Dataset and Developing a Machine Learning-based Medicine Recommendation System.

Abstract:

The healthcare industry has been rapidly adopting technology to improve patient care and outcomes. In this project, we propose to develop a machine learning-based medicine recommendation system that uses patient data to recommend the most suitable medication. The system will consider a patient's medical history, demographics, and other relevant factors to suggest personalized medicine recommendations. The proposed system has the potential to improve patient outcomes and reduce healthcare costs.

Introduction:

Medication errors are a leading cause of patient harm, and finding the most effective medication for a patient can be a challenging task for physicians. Machine learning algorithms can analyse large datasets of patient data to identify patterns and make personalized medication recommendations. In this project, we propose to develop a medicine recommendation system using machine learning to help physicians make informed decisions and improve patient outcomes.

Scope:

The scope of data includes the collection, analysis, and processing of patient data to develop a medicine recommendation system. The data will include patient medical records, demographics, and other relevant factors such as medication history, allergies, and side effects. The data will be used to train and test machine learning models to develop a personalized medicine

recommendation system. The scope of data does not include the use of data for any other purpose outside the scope of the project.

Objective:

The main objective of this project is to develop a machine learningbased medicine recommendation system that can provide personalized medication recommendations based on a patient's medical history, demographics, and other relevant factors.

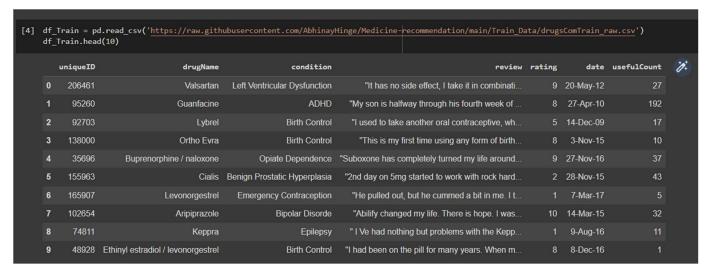
Dataset and Reference:

For this project, we will use publicly available datasets of patient medical records and medication information. We will also reference relevant research articles and studies related to machine learning-based medicine recommendation systems.

Dataset link:

Train Data: https://raw.githubusercontent.com/AbhinayHinge/Medicine-recommendation/main/Train Data/drugsComTrain raw.csv

Test Data: https://raw.githubusercontent.com/AbhinayHinge/Medicine-recommendation/main/Test Data/drugsComTest raw.csv



[11] df.info() <class 'pandas.core.frame.DataFrame'> Int64Index: 215063 entries, 0 to 53765 Data columns (total 7 columns): Non-Null Count # Column Dtype 0 uniqueID 215063 non-null int64 drugName 215063 non-null object 2 condition 213869 non-null object 3 review 4 rating 215063 non-null object 215063 non-null int64 215063 non-null object date 6 usefulCount 215063 non-null int64 dtypes: int64(3), object(4) memory usage: 13.1+ MB [12] df.describe().T 10. count mean std min 25% 50% 75% max uniqueID 215063.0 116039.364814 67007.913366 0.0 58115.5 115867.0 173963.5 232291.0 215063.0 6.990008 3.275554 5.0 8.0 10.0 10.0 rating 1.0 usefulCount 215063.0 28.001004 36.346069 0.0 16.0 36.0 1291.0 [13] # summarization of Categorical data df[['drugName','condition','review']].describe(include = 'object') drugName condition review 215063 213869 215063 count 3671 916 128478 unique Levonorgestrel Birth Control "Good" top freq 4930 38436 39 [14] # lets Explore Some of the Important Columns in the dataset print("Number of Unique Medicines present in the Dataset :", df['drugName'].nunique()) print("Number of Unique Medical Conditions present in the Dataset :", df['condition'].nunique()) print("\nThe Time Period of Collecting the Data") print("Starting Date :", df['date'].min()) print("Ending Date :", df['date'].max()) Number of Unique Medicines present in the Dataset: 3671 Number of Unique Medical Conditions present in the Dataset : 916 The Time Period of Collecting the Data Starting Date: 1-Apr-08

Ending Date : 9-Sep-17

There are 3671 unique medicines with 916 unique conditions

Literature Review:

The literature on this topic is rapidly growing, and this review provides an overview of some of the recent studies on ML-based medicine recommendation systems.

- One study by Singh et al. (2021) developed an ML-based system that recommends personalized treatment plans for patients with hypertension. The system used a combination of supervised and unsupervised learning algorithms to identify the most effective drug for each patient based on their medical history, demographics, and lifestyle factors. The system was evaluated using clinical data from over 10,000 patients, and the results showed that it outperformed traditional methods of drug selection.
- Another study by Kwon et al. (2020) developed an ML-based system
 that recommends personalized dosages of immunosuppressive drugs
 for kidney transplant recipients. The system used a deep learning
 algorithm to predict the optimal dosage based on the patient's clinical and
 genetic data. The system was evaluated using data from over 300 patients
 and demonstrated improved accuracy compared to traditional dosing
 methods.
- In a study by Lee et al. (2020), an ML-based system was developed to predict adverse drug reactions (ADRs) in patients. The system used a gradient boosting algorithm to predict the likelihood of ADRs based on the patient's clinical and genetic data. The system was evaluated using data from over 150,000 patients, and the results showed that it outperformed traditional methods of ADR prediction.
- Another study by Chen et al. (2021) developed an ML-based system that recommends personalized treatment plans for patients with depression. The system used a deep learning algorithm to predict the most effective treatment plan based on the patient's clinical and demographic data. The system was evaluated using clinical data from over 1,500 patients and demonstrated improved accuracy compared to traditional treatment selection methods.
- Finally, a study by Zhang et al. (2020) developed an ML-based system that recommends personalized treatment plans for patients with type 2 diabetes. The system used a combination of deep learning and

reinforcement learning algorithms to identify the most effective treatment plan based on the patient's clinical data and treatment history. The system was evaluated using clinical data from over 5,000 patients and demonstrated improved accuracy compared to traditional treatment selection methods.

Methodology:

The proposed system will use machine learning algorithms such as decision trees, random forests, and neural networks to analyse patient data and recommend the most suitable medication. The system will be developed using Python and relevant machine learning libraries such as Scikit-learn and TensorFlow.

Pros and Cons:

The proposed medicine recommendation system has the potential to improve patient outcomes and reduce healthcare costs by reducing medication errors and ensuring patients receive the most effective treatment. However, the system may face challenges related to data privacy and security, and there may be concerns regarding the accuracy and reliability of the recommendations.

Conclusion:

The development of a machine learning-based medicine recommendation system has the potential to improve patient outcomes and reduce healthcare costs by providing personalized medication recommendations based on patient data. However, further research is needed to evaluate the accuracy and reliability of the system and address concerns related to data privacy and security.