Prediction of Physicians For Patient Diagnosis

Under the guidance of -

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Literature Survey

Paper	Year	Authors	Methodology	Results	Limitations
Disease Prediction and Doctor Recommendation System [1]	March 2018	Tejaswita P. Vaidya Dhanashri Gujar Rashmi Biyani	Predicts disease using Naïve Bayes. Suggesting the details of the disease specialists based on Success rate.	Accuracy above 80 percent. Doctors can be chosen on success rate or fare.	Very few testing and training symptoms and limited filters.
A Hybrid Recommender System for Patient-Doctor Matchmaking in Primary Care [2]	March 2019	Qiwei Han Manas Gaur Leid Zejnilovic	Find doctors visited by similar patients who visits the same doctor.(Hybrid)	Generate a list of doctors for each patient ranked by the predicted trust.	Personalized doctor recommendations.Not for the public(Individual centralized).
Prediction System for Diseases and Suggestion of Appropriate Medicines[3]	December 2017	Disha Mahajan Mrudula Phalak Saniya Pathan	The Classification of data according to requirements and then by applying Association on it for the prediction of diseases	predicting disease and	Not available with doctors (MEDICINES ONLY) and medicine is prescribed for very few diseases.

Paper	Year	Authors	Methodology	Results	Limitations
A Recommender system for identifying the right doctor [4]	December 2016	Li Guo Bo jin Haoyu yang	To develop a recommender system for identifying KOLs for any specific disease with unsupervised learning	recommendation to	Not directly connected to doctors or pharmaceuticals.
Deep Learning Based Health, recommender system using Collaborative Filtering.[5]		Abhaya Sahoo Chittaranjan Pradhan , Rabindra Kumar	This paper gives a proposed Intelligent HRS using Convolutional Neural Network (CNN)	Systems find recommended hospitals by calculating the similarity of patients' choices.	

Outcome of Literature Survey

- The works have been done either with disease prediction or only hospitals recommendation in mind.
- Closely related papers predict a single doctor.
- Most of these papers collaborated with various hospitals and used their data for prediction.
- Hence, the decision was made to take unavailability of doctors into account And thus, the result was to output a list of doctors.

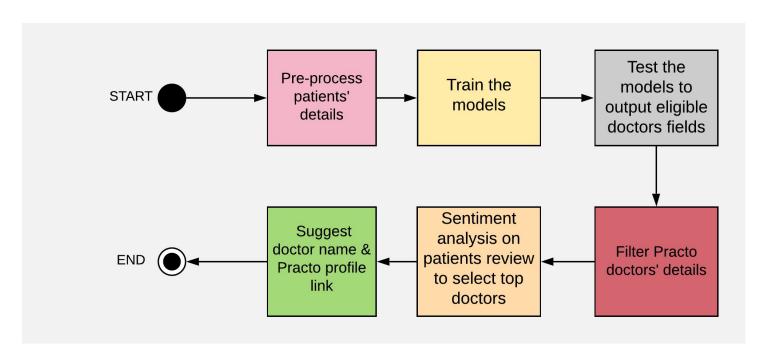
Problem Statement

An efficient predictive model for selecting an appropriate group of healthcare practitioners based on patient details.

Objectives:

- Determining the type of doctor required to diagnose the patient based on patient details i.e. joining two datasets.
- Construct an eligible group of type of doctors to which a patient can refer to, in order.
- Collecting doctors details from open-source data and using a model to construct their rating based on online reviews.
- Mapping the Practo profiles with the predicted type of doctors.

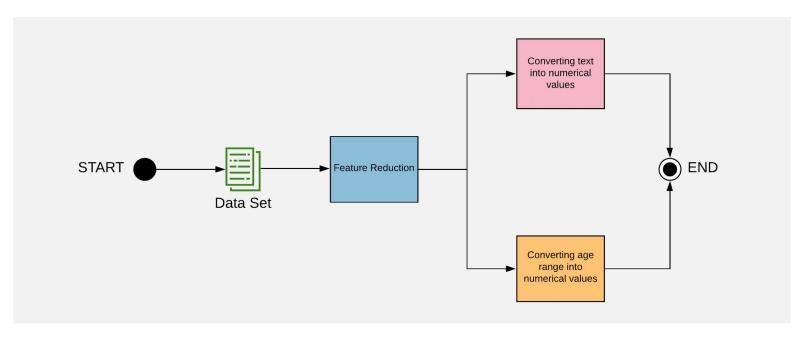
Methodology



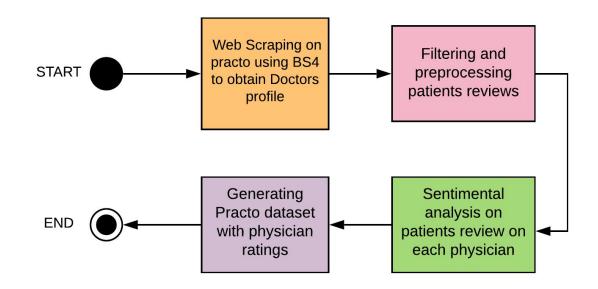
DATASET

- The Health Statewide Planning and Research Cooperative System (SPARCS): It contains Medical information of patient characteristics, diagnoses, treatments and services of patients whose lesional and/or functional status is considered to be stable.
- It has more than 1 lakh observations and 39 variables in its raw state which includes many missing, redundant and irrelevant values. All variables not correlated with the objectives of the study were removed. To solve the problem of missing values, we have used the simple and direct approach that consists of the complete elimination of entries that have missing values.

Patients' Details Pre-processing



Doctors' Reviews Pre-processing



WORK DONE

- Patients' details pre-processing.
- Predicted physicians class using classification models.
- Successfully extracted beta-values.
- Collected doctors profile and ran sentiment analysis on them.
- Implemented a user interface.

Results & Discussion

- Collection of two different datasets was done:
 - Dataset1: After analysing SPARCs dataset we reduced the features.
 - Dataset 2: We created this dataset on our own. It contains 4 symptoms, the disease and the type of doctor required to diagnose the patient.

gastroesophageal reflux disease	pain	pain chest	burning sensation	hyponatremia	gastroenterologist
dehydration	fever	diarrhea	vomiting	hypotension	general physician
cardiomyopathy	shortness of breath	orthopnea	hypokinesia	jugular venous distention	cardiologist
chronic kidney failure	vomiting	orthopnea	hyperkalemia	oliguria	nephrologist
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- Dataset 1 and 2 were combined using a description column, *Diagnosis Description*, in the SPARCS dataset.
- All the textual data was changed to numerical values.
- 11 types of doctors.

18 to 29	F	124 Mir	inor	cough	fever	decreased translucency	shortness of breath	pulmonologist
70 or Older	F	123 Mir	inor	cough	wheezing	shortness of breath	chest tightness	Allergist
70 or Older	M	122 Ma	ajor	wheezing	cough	shortness of breath	chest tightness	Allergist

	Age	Gender	51	52	53	54	Doctor
0	1	1	35	41	11	1	9
1	4	0	7	19	9	41	0
2	3	0	33	35	4	24	0
3	3	1	29	17	48	20	4
4	0	1	19	10	50	25	10

Logistic Regression

- Probability of a certain class or event existing.
- Python's *Sklearn* library.
- Precision and Accuracy for all the 11 classes for logistic regression model using One vs Rest Classifier.
- Store in a Pickle file.

0	0.82	0.86
1	0.82	0.86
2	0.82	0.86
3	0.82	0.86
4	0.83	0.87
5	0.82	0.86
6	0.82	0.86
7	0.87	0.90
8	0.82	0.86
9	0.83	0.87
10	0.83	0.86

Precision Accuracy

- Probabilities of all 11 classes to be able to diagnose the patients for some 3 test cases for logistic regression model using One vs Rest classifier.
- Function called *model.predict_proba()*.

	Allergist	Endocrinologist	General Physician	Cardiologist	Gastroenterologist	Nephrologist	Neurologist	Pediatrician	Psychiatrist	Pulmonologist	Rheumatologist
0	0.905516	0.897717	0.921901	0.913620	0.913422	0.920392	0.897668	0.999573	0.919471	0.912018	0.912724
1	0.912265	0.935740	0.920992	0.939244	0.927671	0.903604	0.899341	0.745244	0.905352	0.933433	0.931883
2	0.881476	0.906254	0.919435	0.912892	0.907052	0.891713	0.883865	0.999990	0.905735	0.908178	0.920634

Random Forest -

- Precision and Accuracy for all the 11 classes for random forest model using One vs Rest Classifier.
- Python's *Sklearn* library.
- Store in a Pickle file.

	PIECISION	Accuracy
0	0.84	0.86
1	0.82	0.86
2	0.82	0.86
3	0.82	0.86
4	0.82	0.86
5	0.82	0.86
6	0.82	0.86
7	0.89	0.90
8	0.82	0.86
9	0.82	0.86
10	0.82	0.86

Precision Accuracy

• Probabilities of all 11 classes to be able to diagnose the patients for some 3 test cases for random forest model using One vs Rest classifier.

	Allergist	Endocrinologist	General Physician	Cardiologist	Gastroenterologist	Nephrologist	Neurologist	Pediatrician	Psychiatrist	Pulmonologist	Rheumatologist
0	0.931355	0.938899	0.933221	0.814891	0.944916	0.934478	0.883007	1.000000	0.932431	0.731708	0.946216
1	0.969033	0.904769	0.898911	0.966588	0.854188	0.905392	0.972800	0.638555	0.937987	1.000000	0.939697
2	0.979752	0.854084	0.855233	0.887698	0.864322	0.849877	0.882552	1.000000	0.958358	0.979421	0.905205

Scraping of doctors' profile details from Practo using BeautifulSoup4.

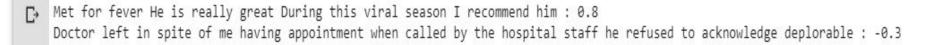
Specialization

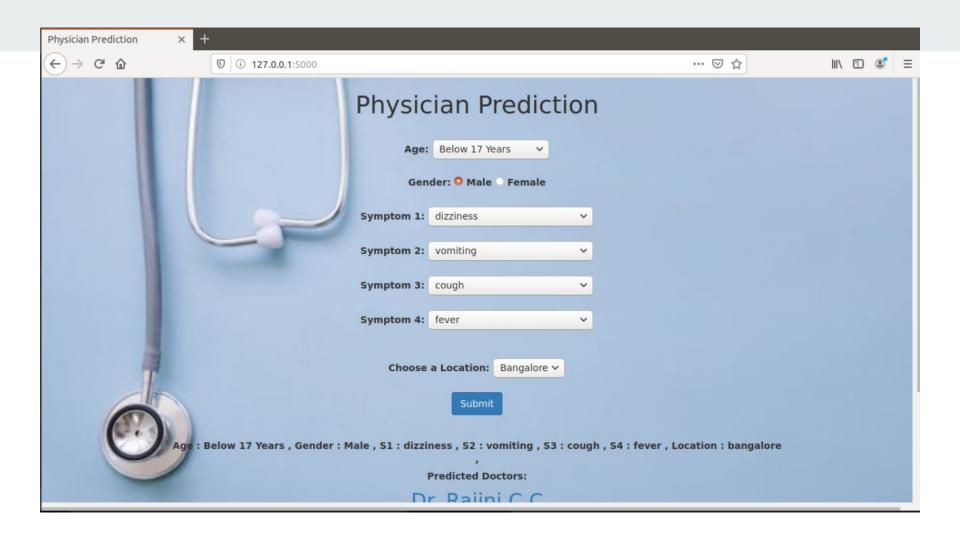
7	Name	Specialization	Location	LITIK
2	Dr. Sheela Chakravarthy	Internal Medicine	bangalore	https://www.practo.com/bangalore/doctor/sheela-chakravarthy-internal-medicine?specialization=Internal Med
3	Dr. Sheetal Kamat	Internal Medicine	bangalore	https://www.practo.com/bangalore/doctor/sheetal-kamat-internal-medicine?specialization=Internal Medicine8
4	Dr. B Rajashekar	General Physician	bangalore	https://www.practo.com/bangalore/doctor/dr-rajashekhar-general-physician?specialization=General Physician&
5	Dr. Raja Selvarajan	Diabetologist	bangalore	https://www.practo.com/bangalore/doctor/dr-raja-selvarajan-diabetologist?specialization=Diabetologist&pract
6	Dr. Sharat Honnatti	general physician	bangalore	https://www.practo.com/bangalore/doctor/dr-sharat-honnatti-general-physician?specialization=&practice_id=
7	Dr. Ashok M N	General Physician	bangalore	https://www.practo.com/bangalore/doctor/dr-ashok-m-n-cardiologist?specialization=General Physician&practi
8	Dr. Tharanath S	General Physician	bangalore	https://www.practo.com/bangalore/doctor/dr-tharanath-s-general-physician?specialization=General Physician&
9	Dr. Pankaj Singhai	Internal Medicine	bangalore	https://www.practo.com/bangalore/doctor/dr-pankaj-singhai-1-internal-medicine?specialization=Internal Medi
10	Dr. Shalini Joshi	Internal Medicine	bangalore	https://www.practo.com/bangalore/doctor/shalini-joshi-internal-medicine?specialization=Internal Medicine≺
11	Dr. Mohan Radagandi	Consent Physician	bangalore	https://www.practo.com/bangaloro/doctor/dr.moban badagandi.diabatalonist2cnosialization=Conocal Dh.wisia

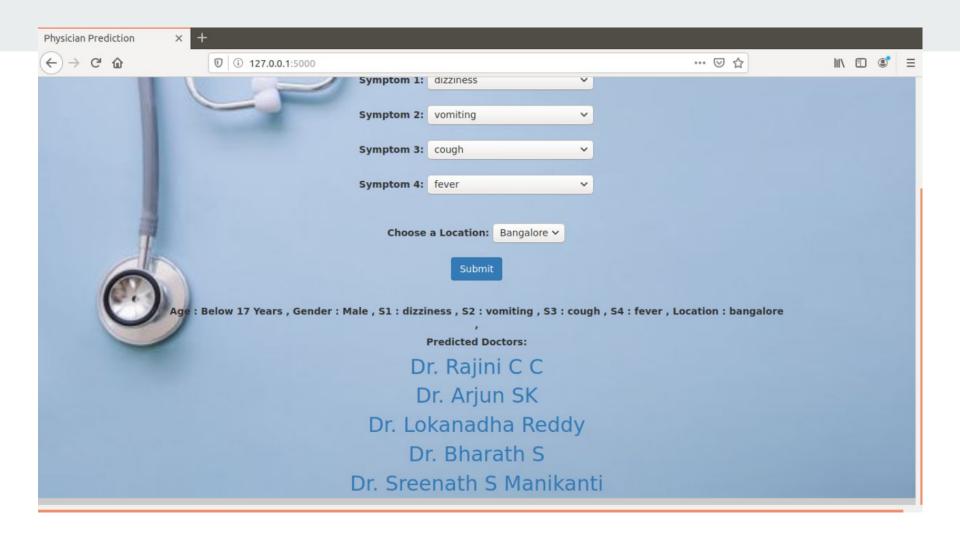
Rating Doctors

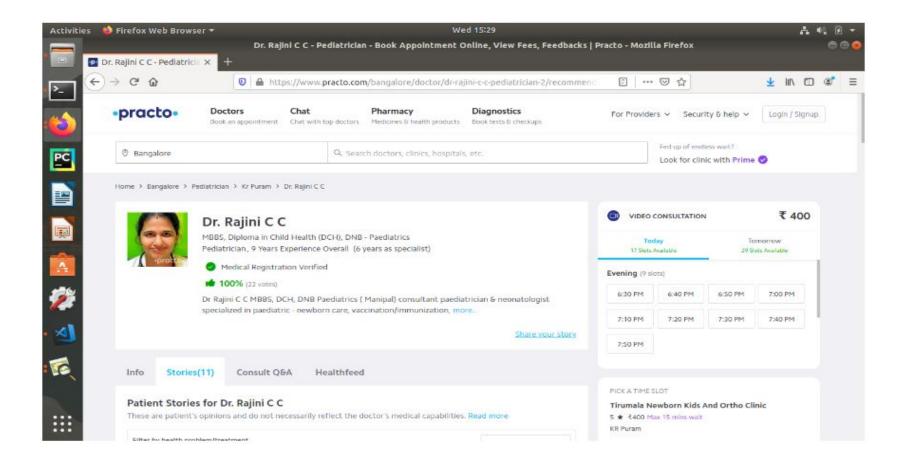
- The sentiment of each each review is calculated using Textblob.
- The Average sentiment of all the reviews converted to percentage is used as doctors rating.

Sentiment value results for some of the reviews:









Individual Contribution

- Bharat Sharma Practo Data Scraping + Data collection + Models + User Interface
- Bharath Simha Pre-processing + Data collection + RF + Textblob
- D. Praneetha Data Mapping + Data collection + LR + Beta Values + Pickle file storage

Conclusion & Future Work

Manual collection of the symptoms, diseases and type of doctors' was done. The two data sets were mapped together and were then pre-processed. Scraping is used for collecting the required Practo data of doctors. The objective to predict the doctor's field is successfully achieved. Sentiment analysis is successfully done on the Practo comments for the respective doctors. And finally, a basic user interface is constructed to bring together both the parts of this project i.e. predicting doctor's field and performing sentiment analysis on Practo data.

The future work is to improve the user interface as well as convert this into an android application. Data from other websites along with Practo can be considered as well. More symptoms and doctor fields can be added.

Timeline



References

- [1] Tejaswita P. Vaidya, Dhanashri Gujar and Rashmi Biyani, Disease Prediction and Doctor Recommendation System.", International Research Journal of Engineering and Technology, Volume 05, Issue 03, March 2018
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www.thebodytransformation.com