TECHNOLOGY BUCKET: HEALTHCARE & BIOMEDICAL DEVICES

COMPANY NAME : ezDI

TEAM NAME : THE_OUTLIERS

TEAM LEADER NAME : APOORVA VIKRAM SINGH

PROBLEM STATEMENT : PATIENT CASE SIMILARITY

PROBLEM CODE : RG2

CATEGORY : **SOFTWARE**

COLLEGE CODE : U-0055

THE SOLUTION:

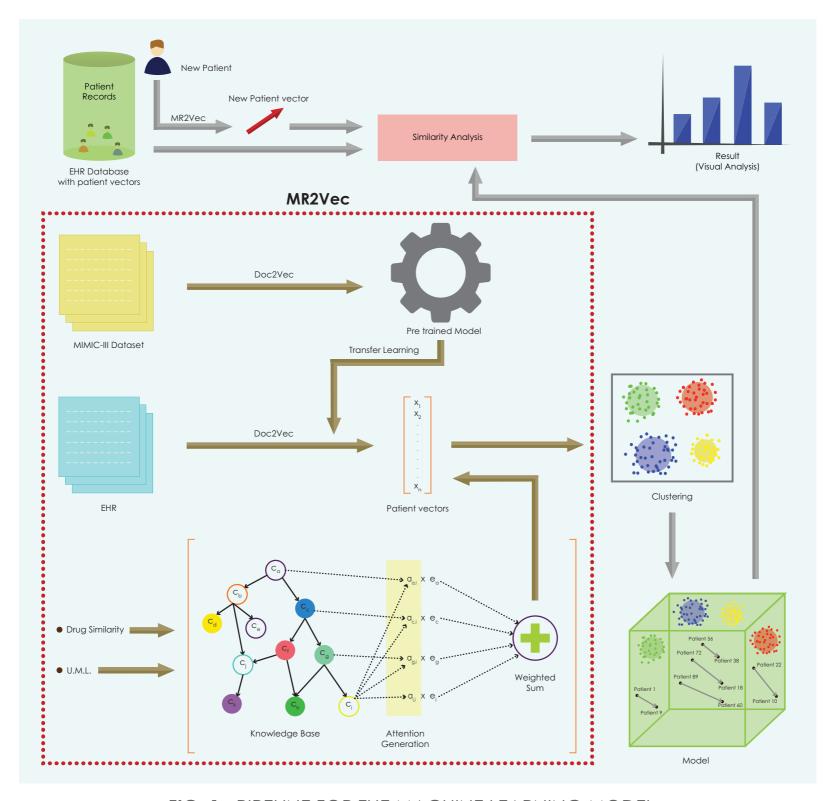
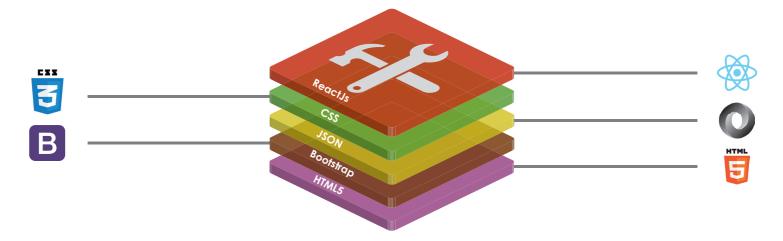


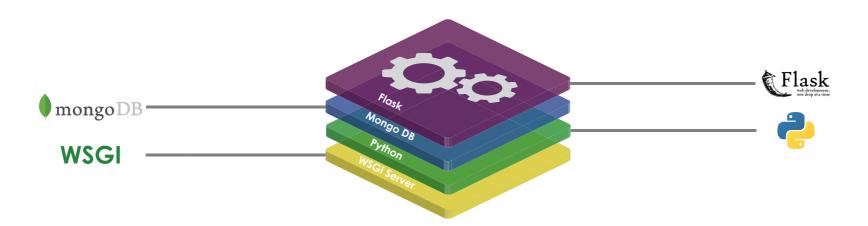
FIG. 1: PIPELINE FOR THE MACHINE LEARNING MODEL.

- MR2Vec: A novel vectorization technique to create patient vectors.
- MR2Vec = Doc2Vec on (MIMIC-III + ezDI) dataset + Transfer Learning + Knowledge
 Base (UMLS database + DrugBank database) + Attention Model
- **Transfer Learning**: To re-train primary model (obtained by deploying Doc2Vec on MIMIC-III) on ezDI dataset to eliminate irrelevant noise.
- **Knowledge Base (KB)**: Assembled by weights obtained from UMLS data and Drug-Bank data, these weights are further multiplied with attention parameters to obtain weighted sums.
- Clustering: Generating patient vector clusters on basis of similarity index.
- **Triplet-Loss Metric**: To calculate normalized distance metric between patient vectors as a measure for similarity score.
- Interactive Interface: Leveraging this machine learning model to engine a website that eases patient similarity analysis practices for Medicare facilities.

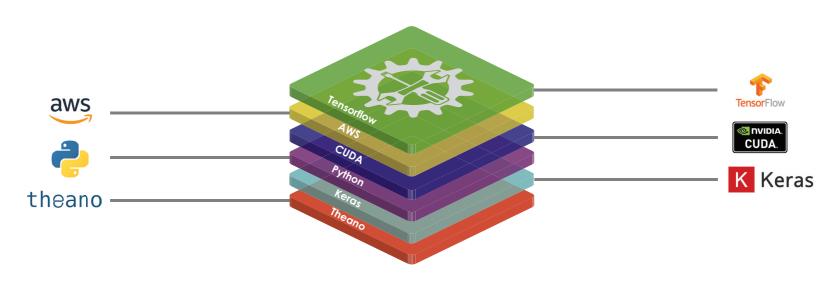
TECHNOLOGY STACK



Client-Side Programming

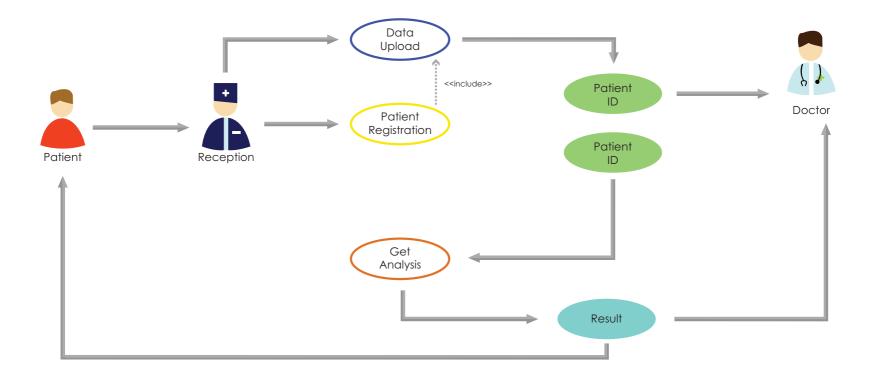


Server-Side Programming

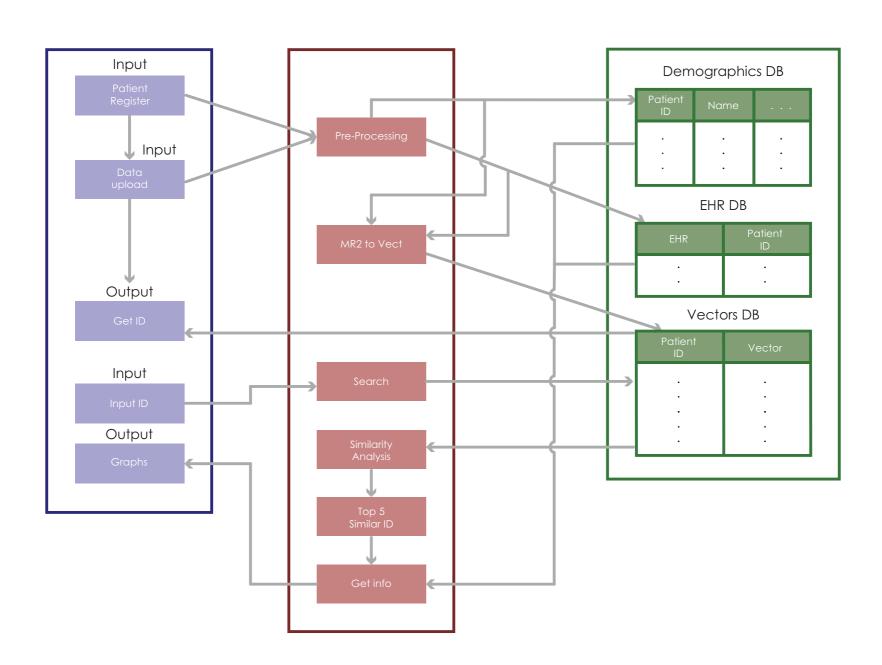


Machine Learning Tools

USE-CASE DIAGRAM



INTERFACE-STRUCTURE DIAGRAM



SHOW-STOPPER

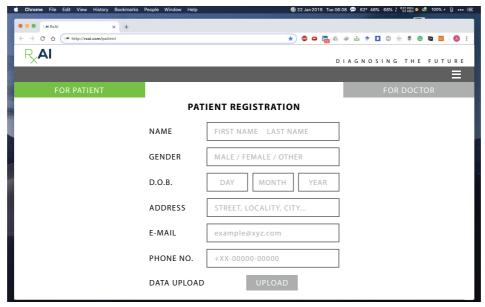


FIG. 2.1: INTERFACE FOR DATA UPLOAD

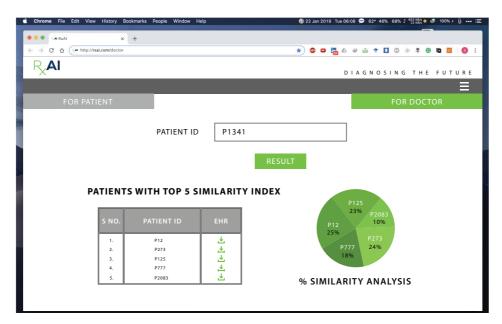


FIG. 2.2: INTERFACE FOR SIMILARITY ANALYSIS (DOCTOR)

- The machine learning model used for predictive analysis of EHR data employs the novel MR2Vec technique for creating vectors of medical records of inpatients.
- MR2Vec utilizes Doc2Vec based intuitive scheme to generate a primary model. This
 primary model is leveraged to train on ezDI dataset to yield a secondary model
 which will be a fine-tuned version of primary model trained on relatively larger MIMIC-III dataset.
- MR2Vec provides much-improved results when compared to generic Word2Vec model. It even provides an intuitive edge over Word2Vec model implemented with pre-trained embeddings which fails to capture thorough semantics from data.
- To further improve the embeddings, we have exploited UMLS database and Drug-Bank to assemble a Knowledge Base (KB) which will yield a weighted sum by multiplying weights with attention parameters.
- We aim to establish an interactive client server interface that proffers a comprehensive anatomization of patient's chronic health and aids Medicare forums in diagnosing the inpatients.