# Custom chatbot predictor for diabetes stages

#### **PROJECT MEMBER DETAILS:**

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## **Responsibilities and Contribution:**

- I had myself contributed 100% to the project from proposal, implementation and to execution
- Using Diabetes symptoms dataset, initially pre processed the dataset and trained three classifiers namely, SVM (Support Vector Machine), K means Clustering, RF (Random Forest)
- Implemented python backend server using Flask
- Implemented chatbot endpoints to send and receive messages
- Did frontend chatbot page in HTML and integrated it with python backend using endpoints via API
- Performed Natural Language Processing and Named Entity Recognition features for the received input messages from the user using Spacy and SciSpacy modules
- Gathered classifier input features from user messages and predicted the diabetes stages

#### **Motivation:**

- Diabetes may not seem harmful initially, but as time goes on, it will affect vital organs in body from heart, vision, kidneys, etc.
- Early stage diabetes prediction is crucial, as proper diagnosis can be provided and in some cases, the diabetes can be reversed without taking any proper medications.
- Currently smart chatbots are the new way of interaction between users and websites
- Currently, there are no smart chat auto assistant which are integrated with machine learning techniques and provide automatic replies
- Chatbot can be easily integrated into any website as plugin which will act as customer support enabled with machine learning assistant

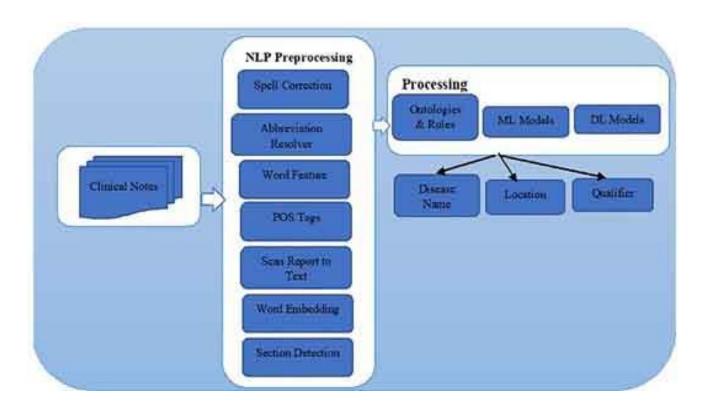
## **Objectives:**

Chatbot should detect entities from user message and provide them to classifier as input features for the classification of diabetes stages

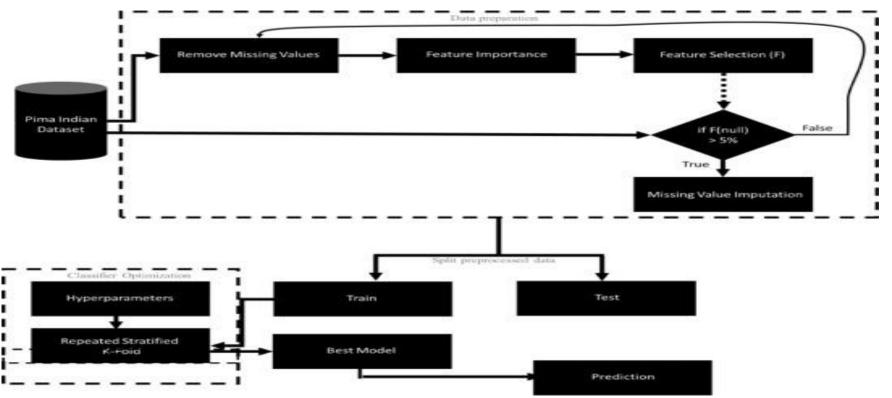
Input can be of two types,

- user message will consists of all symptoms related to diabetes and they will be given as input features to the classifier and the diabetes stage will be predicted.
- User message will be having only one or two symptoms, in this
  case algorithm will prompt questions related to input features
  using chatbot replies and gather respective inputs from user and
  predict diabetes stages.

## **Related Work:** NLP for bio medical terms



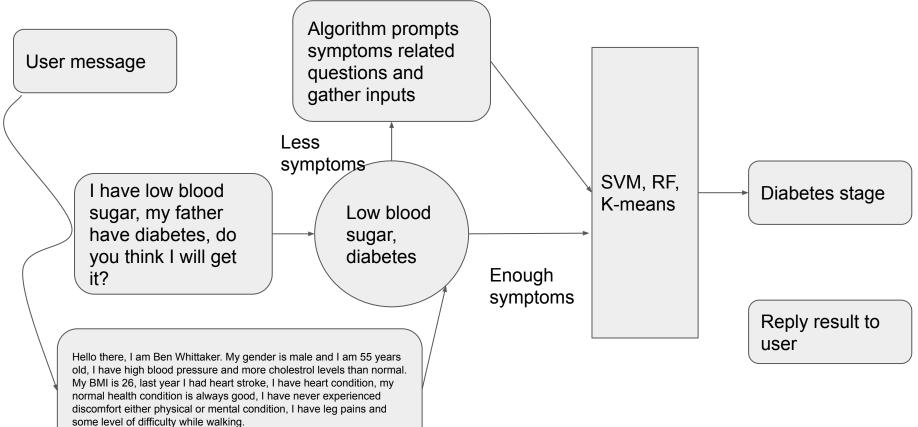
## **Related Work:** Diabetes stages prediction from PIMA dataset



**Problem Statement:** 

Currently there are no chatbot predictors which will gather user inputs and perform classification of diabetes stages

## **Proposed Solution:**



Diabetes	1	0.27	0.21	0.068	0.22	0.063	0.11	0.18	-0.12	-0.042	-0.059	-0.058	0.015	0.035	0.3	0.074	0.18	0.22	0.031	0.19	-0.13	-0.17
HighBP -	0.27	1	0.3	0.099	0.21	0.097	0.13	0.21	-0.13	-0.041	-0.061	-0.004	0.038	0.017	0.3	0.056	0.16	0.22	0.052	0.34	-0.14	-0.17
HighChol -	0.21	0.3	1	0.086	0.11	0.091	0.093	0.18	-0.078	-0.041	-0.04	-0.012	0.042	0.013	0.21	0.062	0.12	0.14	0.031	0.27	-0.071	-0.085
CholCheck -	0.068	0.099	0.086	1	0.034	0.0099	0.024	0.044	0.0042	0.024	0.0061	0.024	0.12	0.058	0.047	0.0084	0.032	0.041	0.022	0.09	0.0015	0.014
BMI -	0.22	0.21	0.11	0.034	1	0.014	0.02	0.053	-0.15	-0.088	-0.062	-0.049	-0.018	0.058	0.24	0.085	0.12	0.2	0.043	-0.037	0.1	-0.1
Smoker -	0.063	0.097	0.091	-0.0099	0.014	1	0.081	0.11	-0.087	-0.078	-0.031	0.1	-0.023	0.049	0.16	0.092	0.12	0.12	0.094	0.12	-0.16	-0.12
Stroke -	0.11	0.13	0.093	0.024	0.02	0.061	1	0.2	-0.069	-0.013	-0.041	-0.017	0.0088	0.035	0.18	0.07	0.15	0.18	0.003	0.13	-0.076	-0.13
HeartDiseaseorAttack -	0.18	0.21	0.18	0.044	0.053	0.11	0.2	1	-0.087	-0.02	-0.039	-0.029	0.019	0.031	0.26	0.065	0.18	0.21	0.086	0.22	-0.1	-0.14
PhysActivity -	-0.12	-0.13	-0.078	0.0042	-0.15	-0.087	-0.069	-0.087	1	0.14	0.15	0.012	0.036	-0.062	-0.27	-0.13	-0.22	-0.25	0.032	-0.093	0.2	0.2
Fruits -	-0.042	0.041	0.041	0.024	-0.088	-0.078	0.013	-0.02	0.14	1	0.25	0.035	0.032	0.044	-0.1	0.068	0.045	0.048	0.091	0.065	0.11	0.08
Veggies -	-0.059	0.061	-0.04	0.0061	-0.062	-0.031	-0.041	-0.039	0.15	0.25	1	0.021	0.03	-0.032	-0.12	-0.059	-0.064	-0.081	-0.065	-0.0098	0.15	0.15
HvyAlcoholConsump -	-0.058	-0.004	-0.012	-0.024	-0.049	0.1	-0.017	-0.029	0.012	-0.035	0.021	1	-0.01	0.0047	-0.037	0.025	-0.026	-0.038	0.0057	-0.035	0.024	0.054
AnyHealthcare –	0.015	0.038	0.042	0.12	-0.018	-0.023	0.0088	0.019	0.036	0.032	0.03	-0.01	1	-0.23	-0.041	-0.053	-0.0083	0.0071	-0.019	0.14	0.12	0.16
NoDocbcCost -	0.035	0.017	0.013	-0.058	0.058	0.049	0.035	0.031	-0.062	-0.044	-0.032	0.0047	-0.23	1	0.17	0.19	0.15	0.12	-0.045	-0.12	-0.1	-0.2
GenHlth -	0.3	0.3	0.21	0.047	0.24	0.16	0.18	0.26	-0.27	-0.1	-0.12	-0.037	-0.041	0.17	1	0.3	0.52	0,46	-0.0061	0.15	-0.28	-0.37
MentHith -	0.074	0.056	0.062	0.0084	0.085	0.092	0.07	0.065	0.13	-0.068	0.059	0.025	0.053	0.19	0.3	1	0.35	0.23	0.081	-0.092	-0.1	-0.21
PhysHith -	0.18	0.16	0.12	0.032	0.12	0.12	0.15	0.18	-0.22	-0.045	-0.064	-0.026	-0.0083	0.15	0.52	0.35	1	0.48	-0.043	0.099	0.16	-0.27
DiffWalk -	0.22	0.22	0.14	0.041	0.2	0.12	0.18	0.21	-0.25	-0.048	-0.081	-0.038	0.0071	0.12	0.46	0.23	0.48	1	-0.07	0.2	-0.19	-0.32
Sex -	0.031	0.052	0.031	-0.022	0.043	0.094	0.003	0.086	0.032	-0.091	-0.065	0.0057	-0.019	-0.045	-0.0061	-0.081	-0.043	-0.07	1	-0.027	0.019	0.13
Age -	0.19	0.34	0.27	0.09	-0.037	0.12	0.13	0.22	-0.093	0.065	-0.0098	-0.035	0.14	-0.12	0.15	-0.092	0.099	0.2	-0.027	1	-0.1	-0.13
Education -	-0.13	-0.14	-0.071	0.0015	-0.1	-0.16	-0.076	-0.1	0.2	0.11	0.15	0.024	0.12	-0.1	-0.28	-0.1	-0.16	-0.19	0.019	-0.1	1	0.45
Income -	-0.17	-0.17	0.085	0.014	-0.1	-0.12	0.13	0.14	0.2	0.08	0.15	0.054	0.16	-0.2	0.37	0.21	0.27	-0.32	0.13	0.13	0.45	1
	Diabetes -	HighBP	HighChol	chalcheck.	BIMI -	Smoker	Stroke	HeartDiseaseorAttack	PhysActivity -	Pruits	Veggies	HvyAlcoholConsump	AnyHealthcare	MoDocbcCost	GenHith.	MentHith	PhysHlth -	Diffwalk	3	Vge	Education	Income

0.75

- 0.50

0.25

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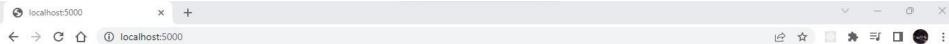
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-0.75

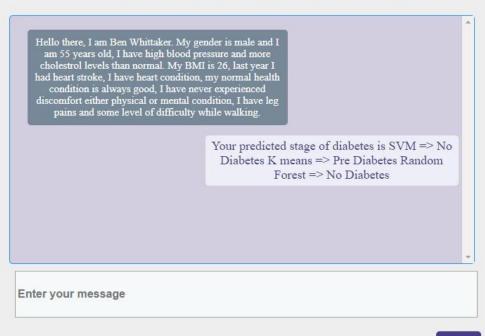
-1.00

22855	0	0	1	7	
121922	30	1	1	10	# Random Forest
232355	0	0	0	9	from sklearn.ensemble import RandomForestClassifier
208496	10	0	0	7	<pre>rf model = RandomForestClassifier(n estimators=100, max features="auto", random state=44)</pre>
232431	3	0	0	10	rf_model.fit(X_train, y_train)
[50736 rd	ows x 10	columns]			rf_pred = rf_model.predict(X_test)
Fitting (					<pre>print("RF predicted : \n", rf_pred)</pre>
SVC Class	sificatio	n Report:			from sklearn import metrics
	pr	ecision	recal	l f1-sco	
	0	0.84	1.00	0.9	
	1	0.00	0.00	0.0	pickle.dump(rf_model, open('rf_model.pkl', 'wb'))
	2	0.62	0.02	0.0	
accur	racy			0.8	RF predicted :
macro	avg	0.49	0.34	0.3	[0 0 0 0 0 2]
weighted	avg	0.80	0.84	0.7	Accuracy: 0.826631977294229
Score: (	8442673	841059603			

# Results



#### Custom Chatbot Predictor For Diabetes Stages

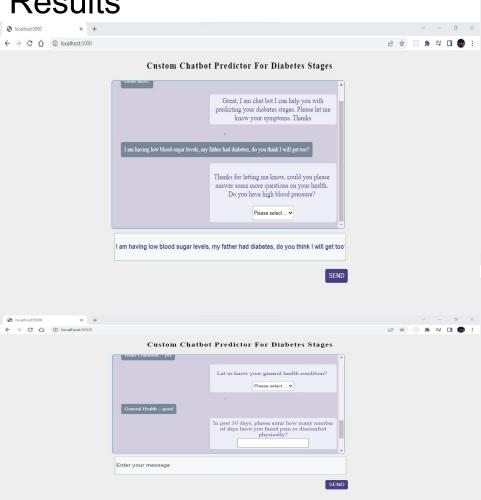


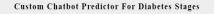
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# Results

```
C:\WINDOWS\system32\cmd.exe - flask run
127.0.0.1 - - [05/Dec/2022 19:29:43] "GET / HTTP/1.1" 200 -
received message Hello there, I am Ben Whittaker. My gender is male and I am 55 years old, I have high blood pressure and more cholestrol levels than normal. My BMI is
26, last year I had heart stroke, I have heart condition, my normal health condition is always good, I have never experienced discomfort either physical or mental condi
tion. I have leg pains and some level of difficulty while walking.
General entities Ben Whittaker 18 31 PERSON
General entities 55 years old 60 72 DATE
General entities BMI 144 147 ORG
General entities 26 151 153 DATE
General entities last year 155 164 DATE
Hello 0 5 ENTITY
gender 36 42 ENTITY
male 46 50 ENTITY
I 55 56 ENTITY
years 63 68 ENTITY
I 74 75 ENTITY
blood pressure 86 100 ENTITY
cholestrol 110 120 ENTITY
levels 121 127 ENTITY
normal 133 139 ENTITY
BMI 144 147 ENTITY
vear 160 164 ENTITY
heart stroke 171 183 ENTITY
I 185 186 ENTITY
heart condition 192 207 ENTITY
health condition 219 235 ENTITY
I 252 253 ENTITY
discomfort 277 287 ENTITY
physical 295 303 ENTITY
mental condition 307 323 ENTITY
I 325 326 ENTITY
leg pains 332 341 ENTITY
level 351 356 ENTITY
difficulty 360 370 ENTITY
walking 377 384 ENTITY
Entities: ['Ben Whittaker', '55 years old', 'BMI', '26', 'last year', 'Hello', 'gender', 'male', 'I', 'years', 'I', 'blood pressure', 'cholestrol', 'levels', 'normal',
'BMI', 'year', 'heart stroke', 'I', 'heart condition', 'health condition', 'I', 'discomfort', 'physical', 'mental condition', 'I', 'leg pains', 'level', 'difficulty',
walking'l
11 output
Array from entities: [[ 1  1  25  1  1  1  20  1  0  40]]
Inputs from entities : [[ 1  1  25  1  1  1  20  1  0  40]]
got output : {'svm_predicted': {'output': 'No Diabetes'}, 'km_predicted': {'output': 'Pre Diabetes'}, 'rf_predicted': {'output': 'No Diabetes'}}
127.0.0.1 - - [05/Dec/2022 19:29:47] "POST /send message HTTP/1.1" 200 -
```

## Results





O localhost:5000

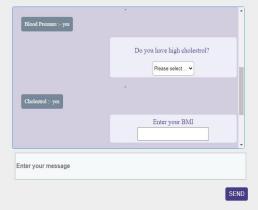
O localhost:5000

← → C ☆ ① localhost:5000

← → C ♠ ① localhost:5000

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#### Custom Chatbot Predictor For Diabetes Stages



#### **REFERENCES:**

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- Oyebode, O., Alqahtani, F., & Orji, R. (2020). Using Machine Learning and Thematic Analysis Methods to Evaluate Mental Health Apps Based on User Reviews. IEEE, 18.

# THANK YOU