

# **HEALTH CARE CHAT BOT**

## **A MINI PROJECT REPORT**

**18CSC305J - ARTIFICIAL INTELLIGENCE**

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# **SRM INSTITUTE OF SCIENCE AND TECHNOLOGY**

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## **BONAFIDE CERTIFICATE**

Certified that the Mini project report titled “**HEALTH CARE CHAT BOT**” is the bona fide work of **BUCHINGARI VASU (RA2011003010412), BALA CHANDRA KARNA(RA2011003010435), SHAIK HUSSAIN AHAMED(RA2011003010439)** who carried out the minor project under my supervision. Certified further, that to the best of my knowledge, the work reported herein does not form any other project report or dissertation on the basis of which a degree or award was conferred on an earlier occasion on this or any other candidate.

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## **ABSTRACT**

The Health Care Chatbot AI project aims to develop an intelligent conversational agent that can provide healthcare information and support to users through natural language interactions. The chatbot will utilize machine learning algorithms and NLP techniques to understand user inputs and provide relevant and accurate responses. The system will be designed to handle a wide range of healthcare-related queries, including symptoms, treatments, medications, and medical conditions. The project will leverage existing healthcare datasets to improve the chatbot's accuracy and relevance, while also prioritizing user privacy and data security. Ultimately, the Health Care Chatbot AI project aims to provide users with an accessible and convenient means of accessing reliable healthcare information and support.

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**ABSTRACT**

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# **CHAPTER 1**

## **INTRODUCTION**

A computer programme created for the healthcare sector's chat bot initiative simulates talks with real consumers. The chat bot intends to help patients and healthcare professionals by giving medical advice, responding to inquiries about symptoms and treatments, directing users to resources for healthcare, and helping with appointment scheduling. The need for rapid and easy access to medical information as well as the rising demand for healthcare services have increased the significance of the health chat bot project. Users can communicate with the healthcare system through chat bots, which provide an easy-to-use and accessible method of communication. The health care chat bot project makes use of a number of technologies, including artificial intelligence, machine learning, and natural language processing, to comprehend and reply to user requests. By analysing user inputs and delivering pertinent information, the chatbot can provide precise and prompt responses. Overall, the health care chat bot project is a useful tool for the healthcare sector, offering a low-cost and effective way to provide consumers and healthcare providers with medical advice and support.

## **CHAPTER 2**

### **LITERATURE SURVEY**

In recent years, the use of chatbots in the healthcare industry has increased. A number of research have been carried out to determine how useful chatbots are in various healthcare settings.

Title of Paper 1: Improving LMS Experience with AIML Base and Retrieval Base R-based chatbot Language

Brief Description: The usage of several techniques, including N-gram, Stemming, TF-IDF, and cosine similarity, is covered in detail in the study. This paper explains how to apply these techniques to produce an optimised outcome rapidly. It details various inquiries and how a chatbot might respond to them. For convenience, they also specified the functional architecture of databases and offered test cases for algorithms. The essence of chatbot quality

Title of Paper 2: Artificial Intelligence Based Personal Assistant,

Brief Description:

Here In this work, a chatbot that acts as a user's personal healthcare assistant is built for healthcare- related purposes. A user can interact with a medical assistant via a dialogue interface. It offers features like ailment diagnosis based on symptoms reported by the user, definitions of medical terms, advice from doctors, scheduling of treatments, and tracking and monitoring of the user's health metrics. Both offline and online performance were looked at overall. To assess the effectiveness and quality of the system, they also ran a number of tests.

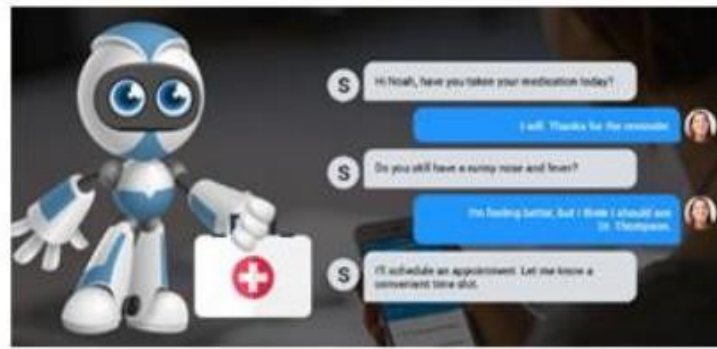
To create an eHealth environment for patients' convenience, numerous studies on this topic have been done in the recent years.

Title of Paper 3: Human-to-Machine Conversation Modelling

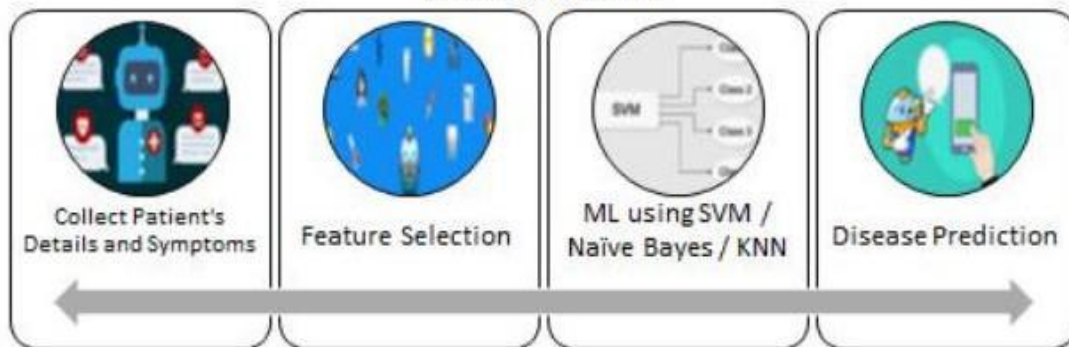
Brief Description: In this study, hardware and software were combined to develop a chatbot system. This chatbot is Bluetooth-enabled and can move. They controlled chatbots through voice communication. A database that is kept on the Raspberry Pi will be searched using the user's query. This chatbot only functions with particular illnesses like the common cold, typhoid, malaria, etc.

# CHAPTER 3

## SYSTEM ARCHITECTURE AND DESIGN



**Medical Chabot**



This is the proposed architecture for our AI-Powered Health Chatbot, and we will give further explain for every component in the next upcoming papers. First of all, the end user interacts with the chatbot through a client platform. It's important to be user friendly and give an excellent UX.

Each time, the user is having a request, it's routed to the NLP Engine using the appropriate API's, in the NLP Engine, and with NLU, the chatbot understands the request and format the data into understandable form that can be understood by the Core Engine. Once the Core Engine receives the formatted data, it searches using Deep Learning Algorithms, the appropriate response and send it back to the NLP Engine.

## **CHAPTER 4**

### **METHODOLOGY**

The following steps could be used in the technique for creating a health care chatbot:

A health care chatbot's target audience must be determined before any other steps can be taken to develop it. The chatbot might be created to meet the needs of patients, physicians, nurses, or other healthcare providers. Designing the chatbot's functionalities would be aided by an understanding of the needs of the target audience.

**Identify the chatbot's objectives:** The next stage is to identify the objectives of the chatbot. The objectives should match the requirements of the target audience. For instance, a chatbot created just for patients would be intended to answer general health-related concerns, arrange appointments, and send out reminders.

**Gather information and create a database:** A substantial amount of data that can aid the chatbot in comprehending and responding to user inquiries is necessary in order to construct one. The information might be found in patient records, clinical data, and medical literature. By using this information to teach the chatbot, a knowledge base can be created.

**Select the chatbot development platform and tools:** There are a variety of chatbot development platforms and tools available. The objectives and needs of the chatbot will determine the platform and technologies to use.

**Create the dialogue flow for the chatbot:** This will establish how the chatbot will communicate with the user. The discussion flow has to be planned to guarantee that the user's inquiries are answered truthfully and effectively.

**Create the natural language processing (NLP) system for the chatbot:** The chatbot's capacity to comprehend natural language is a crucial feature. The NLP system needs to be able to handle many dialects, accents, and languages.

**Test and improve the chatbot:** After the chatbot is created, it should be put to the test to make sure it works as intended. Any problems or flaws with the chatbot can be found by conducting user testing. The chatbot can be enhanced and developed based on the input.

**Deploy the chatbot:** After it has been tested and improved, the chatbot can be made available to the intended audience. To make sure the chatbot is operating properly and answering users' questions, regular checks should be made.



# CHAPTER 5

## CODING AND TESTING

```

1 ~ import re
2 import pandas as pd
3 import pyttx3
4 from sklearn import preprocessing
5 from sklearn.tree import DecisionTreeClassifier, _tree
6 import numpy as np
7 from sklearn.model_selection import train_test_split
8 from sklearn.model_selection import cross_val_score
9 from sklearn.svm import SVC
10 import csv
11 import warnings
12 warnings.filterwarnings("ignore", category=DeprecationWarning)
13
14
15 training = pd.read_csv('Data/Training.csv')
16 testings = pd.read_csv('Data/Testing.csv')
17 cols = training.columns
18 cols = cols[:-1]
19 x = training[cols]
20 y = training['prognosis']
21 y1 = y
22
23
24 reduced_data = training.groupby(training['prognosis']).max()
25
26 #mapping strings to numbers
27 le = preprocessing.LabelEncoder()
28 le.fit(y)
29 y = le.transform(y)

```

```

12 x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.33, random_state=42)
13 testx = testing[cols]
14 testy = testing['prognosis']
15 testy = le.transform(testy)
16 clf1 = DecisionTreeClassifier()
17 clf = clf1.fit(x_train, y_train)
18 # print(clf.score(x_train, y_train))
19 # print ("cross result=====")
20 scores = cross_val_score(clf, x_test, y_test, cv=3)
21 # print (scores)
22 print (scores.mean())
23 model=SVC()
24 model.fit(x_train, y_train)
25 print("for svm: ")
26 print(model.score(x_test, y_test))
27 importances = clf.feature_importances_
28 indices = np.argsort(importances)[::-1]
29 features = cols
30 def readn(nstr):
31     engine = pyttx3.init()
32     engine.setProperty('voice', "english+f5")
33     engine.setProperty('rate', 150)
34     engine.say(nstr)
35     engine.runAndWait()
36     engine.stop()
37 severityDictionary=dict()
38 description_list = dict()
39 precautionDictionary=dict()
40 symptoms_dict = {}

```

```

62 for index, symptom in enumerate(x):
63     symptoms_dict[symptom] = index
64
65 1 usage:
66 def calc_condition(exp, days):
67     sum=0
68     for item in exp:
69         sum=sum+severityDictionary[item]
70     if((sum*days)/(len(exp)+1)>13):
71         print("You should take the consultation from doctor. ")
72     else:
73         print("It might not be that bad but you should take precautions.")
74
75 1 usage:
76 def getDescription():
77     global description_list
78     with open('MasterData/symptom_Description.csv') as csv_file:
79         csv_reader = csv.reader(csv_file, delimiter=',')
80         line_count = 0
81         for row in csv_reader:
82             _description={row[0]:row[1]}
83             description_list.update(_description)
84
85 1 usage:
86 def getSeverityDict():
87     global severityDictionary
88     with open('MasterData/symptom_severity.csv') as csv_file:
89
90         csv_reader = csv.reader(csv_file, delimiter=',')
91         line_count = 0

```

```

88 try:
89     for row in csv_reader:
90         _diction={row[0]:int(row[1])}
91         severityDictionary.update(_diction)
92     except:
93         pass
94
95 1 usage:
96 def getprecautionDict():
97     global precautionDictionary
98     with open('MasterData/symptom_precaution.csv') as csv_file:
99
100         csv_reader = csv.reader(csv_file, delimiter=',')
101         line_count = 0
102         for row in csv_reader:
103             _precc={row[0]:[row[1],row[2],row[3],row[4]]}
104             precautionDictionary.update(_precc)
105
106 1 usage:
107 def getInfo():
108     print("-----HealthCare ChatBot-----")
109     print("\nYour Name? \t\t\t\t\t,ends=->")
110     name=input("")
111     print("Hello, ", name)
112
113 1 usage:
114 def check_pattern(dis_list, inp):

```

```

113 def check_pattern(dis_list,inp):
114     mod_lists=[]
115     inp=inp.replace(' ','_')
116     patt = f"^{inp}$"
117     regexp = re.compile(patt)
118     pred_list=[item for item in dis_list if regexp.search(item)]
119     if(len(pred_list)>0):
120         return 1,pred_list
121     else:
122         return 0,[]
123
124 def sec_predict(symptoms_exp):
125     df = pd.read_csv('Data/Training.csv')
126     X = df.iloc[:, :-1]
127     y = df['prognosis']
128     X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=20)
129     rf_clf = DecisionTreeClassifier()
130     rf_clf.fit(X_train, y_train)
131
132     symptoms_dict = {symptom: index for index, symptom in enumerate(X)}
133     input_vector = np.zeros(len(symptoms_dict))
134     for item in symptoms_exp:
135         input_vector[symptoms_dict[item]] = 1
136
137     return rf_clf.predict([input_vector])
138
139 def print_disease(node):

```

```

168 conf_inp=0
169
170 disease_input=cnf_dis[conf_inp]
171 break
172 # print("Did you mean: ",cnf_dis,"?(yes/no) :",end="")
173 # conf_inp = input("")
174 # if(conf_inp=="yes"):
175 #     break
176 else:
177     print("Enter valid symptom.")
178
179 while True:
180     try:
181         num_days=int(input("Okay. From how many days ? : "))
182         break
183     except:
184         print("Enter valid input.")
185 def recurse(node, depth):
186     indent = " " * depth
187     if tree_.feature[node] != _tree.TREE_UNDEFINED:
188         name = feature_name[node]
189         threshold = tree_.threshold[node]
190
191         if name == disease_input:
192             val = 1
193         else:
194             val = 0
195         if val <= threshold:
196             recurse(tree_.children_left[node], depth + 1)

```

```

198 symptoms_present.append(name)
199 recurse(tree_.children_right[node], depth + 1)
200 else:
201     present_disease = print_disease(tree_.value[node])
202     # print("You may have " + present_disease )
203     red_cols = reduced_data.columns
204     symptoms_given = red_cols[reduced_data.loc[present_disease].values[0].nonzero()]
205     # dis_list=list(symptoms_present)
206     # if len(dis_list)!=0:
207     #     print("symptoms present " + str(list(symptoms_present)))
208     # print("symptoms given " + str(list(symptoms_given)) )
209     print("Are you experiencing any ")
210     symptoms_exp=[]
211     for syms in list(symptoms_given):
212         inp=""
213         print(syms,"? : ",end='')
214         while True:
215             inp=input("")
216             if(inp=="yes" or inp=="no"):
217                 break
218             else:
219                 print("provide proper answers i.e. (yes/no) : ",end="")
220             if(inp=="yes"):
221                 symptoms_exp.append(syms)
222
223 second_prediction=sec_predict(symptoms_exp)
224 # print(second_prediction)
225 calc_condition(symptoms_exp,num_days)
226 if(present_disease[0]==second_prediction[0]):
227     # print("You may have " + present_disease[0])
228     # print(description_list[present_disease[0]])
229     # readn(f"You may have {present_disease[0]}")
230     # readn(f"{description_list[present_disease[0]]}")
231
232 else:
233     print("You may have ", present_disease[0], "or ", second_prediction[0])
234     print(description_list[present_disease[0]])
235     print(description_list[second_prediction[0]])
236
237 # print(description_list[present_disease[0]])
238 precaution_list=precautionDictionary[present_disease[0]]
239 print("Take following measures : ")
240 for i,j in enumerate(precaution_list):
241     print(i+1,".",j)
242
243 # confidence_level = (1.0*(len(symptoms_present))/len(symptoms_given))
244 # print("confidence level is " + str(confidence_level))
245
246 recurse(0, 1)
247 getSeverityDict()
248 getDescription()
249 getPrecautionDict()
250 getInfo()
251 tree_to_code(clf,cols)
252 print("-----")

```

```

253 calc_condition(symptoms_exp,num_days)
254
255 if(present_disease[0]==second_prediction[0]):
256     print("You may have ", present_disease[0])
257     print(description_list[present_disease[0]])
258
259     # readn(f"You may have {present_disease[0]}")
260     # readn(f"{description_list[present_disease[0]]}")
261
262 else:
263     print("You may have ", present_disease[0], "or ", second_prediction[0])
264     print(description_list[present_disease[0]])
265     print(description_list[second_prediction[0]])
266
267 # print(description_list[present_disease[0]])
268 precaution_list=precautionDictionary[present_disease[0]]
269 print("Take following measures : ")
270 for i,j in enumerate(precaution_list):
271     print(i+1,".",j)
272
273 # confidence_level = (1.0*(len(symptoms_present))/len(symptoms_given))
274 # print("confidence level is " + str(confidence_level))
275
276 recurse(0, 1)
277 getSeverityDict()
278 getDescription()
279 getPrecautionDict()
280 getInfo()
281 tree_to_code(clf,cols)
282 print("-----")

```

## Code:

```
import re
import pandas as pd
import pyttsx3
from sklearn import preprocessing
from sklearn.tree import DecisionTreeClassifier, _tree
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.model_selection import cross_val_score
from sklearn.svm import SVC
import csv
import warnings
warnings.filterwarnings("ignore", category=DeprecationWarning)

training = pd.read_csv('Data/Training.csv')
testing = pd.read_csv('Data/Testing.csv')
cols = training.columns
cols = cols[:-1]
x = training[cols]
y = training['prognosis']
y1 = y

reduced_data = training.groupby(training['prognosis']).max()

#mapping strings to numbers
le = preprocessing.LabelEncoder()
le.fit(y)
y = le.transform(y)

x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.33, random_state=42)
testx = testing[cols]
testy = testing['prognosis']
testy = le.transform(testy)

clf1 = DecisionTreeClassifier()
clf = clf1.fit(x_train, y_train)
# print(clf.score(x_train, y_train))
# print ("cross result=====")
scores = cross_val_score(clf, x_test, y_test, cv=3)
# print (scores)
print (scores.mean())

model = SVC()
model.fit(x_train, y_train)
print("for svm: ")
print(model.score(x_test, y_test))

importances = clf.feature_importances_
indices = np.argsort(importances)[::-1]
features = cols

def readn(nstr):
    engine = pyttsx3.init()

    engine.setProperty('voice', 'english+f5")
    engine.setProperty('rate', 130)

    engine.say(nstr)
    engine.runAndWait()
    engine.stop()

severityDictionary=dict()
description_list = dict()
precautionDictionary=dict()

symptoms_dict = { }
```

```

for index, symptom in enumerate(x):
    symptoms_dict[symptom] = index
def calc_condition(exp,days):
    sum=0
    for item in exp:
        sum=sum+severityDictionary[item]
    if((sum*days)/(len(exp)+1)>13):
        print("You should take the consultation from doctor. ")
    else:
        print("It might not be that bad but you should take precautions.")

def getDescription():
    global description_list
    with open('MasterData/symptom_Description.csv') as csv_file:
        csv_reader = csv.reader(csv_file, delimiter=',')
        line_count = 0
        for row in csv_reader:
            _description={row[0]:row[1]}
            description_list.update(_description)

def getSeverityDict():
    global severityDictionary
    with open('MasterData/symptom_severity.csv') as csv_file:

        csv_reader = csv.reader(csv_file, delimiter=',')
        line_count = 0
        try:
            for row in csv_reader:
                _diction={row[0]:int(row[1])}
                severityDictionary.update(_diction)
        except:
            pass

def getprecautionDict():
    global precautionDictionary
    with open('MasterData/symptom_precaution.csv') as csv_file:

        csv_reader = csv.reader(csv_file, delimiter=',')
        line_count = 0
        for row in csv_reader:
            _prec={row[0]:[row[1],row[2],row[3],row[4]]}
            precautionDictionary.update(_prec)

def getInfo():
    print("-----HealthCare ChatBot-----")
    print("\nYour Name? \t\t\t",end="->")
    name=input("")
    print("Hello, ",name)

def check_pattern(dis_list,inp):
    pred_list=[]
    inp=inp.replace(' ','_')
    patt = f"{inp}"
    regexp = re.compile(patt)
    pred_list=[item for item in dis_list if regexp.search(item)]
    if(len(pred_list)>0):
        return 1,pred_list
    else:
        return 0,[]

def sec_predict(symptoms_exp):
    df = pd.read_csv('Data/Training.csv')
    X = df.iloc[:, :-1]
    y = df['prognosis']
    X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=20)
    rf_clf = DecisionTreeClassifier()
    rf_clf.fit(X_train, y_train)

```

```

symptoms_dict = {symptom: index for index, symptom in enumerate(X)}
input_vector = np.zeros(len(symptoms_dict))
for item in symptoms_exp:
    input_vector[[symptoms_dict[item]]] = 1

return rf_clf.predict([input_vector])

def print_disease(node):
    node = node[0]
    val = node.nonzero()
    disease = le.inverse_transform(val[0])
    return list(map(lambda x:x.strip(),list(disease)))

def tree_to_code(tree, feature_names):
    tree_ = tree.tree_
    feature_name = [
        feature_names[i] if i != _tree.TREE_UNDEFINED else "undefined!"
        for i in tree_.feature
    ]

    chk_dis=", ".join(feature_names).split(",")
    symptoms_present = []

    while True:

        print("\nEnter the symptom you are experiencing \t\t,end="->")
        disease_input = input("")
        cnf,cnf_dis=check_pattern(chk_dis,disease_input)
        if cnf==1:
            print("searches related to input: ")
            for num,it in enumerate(cnf_dis):
                print(num,"") ,it)
            if num!=0:
                print(f"Select the one you meant (0 - {num}): ", end="")
                conf_inp = int(input(""))
            else:
                conf_inp=0

            disease_input=cnf_dis[conf_inp]
            break
            # print("Did you mean: ",cnf_dis,"?(yes/no) :",end="")
            # conf_inp = input("")
            # if(conf_inp=="yes"):
            #     break
        else:
            print("Enter valid symptom.")

    while True:
        try:
            num_days=int(input("Okay. From how many days ? : "))
            break
        except:
            print("Enter valid input.")
    def recurse(node, depth):
        indent = " " * depth
        if tree_.feature[node] != _tree.TREE_UNDEFINED:
            name = feature_name[node]
            threshold = tree_.threshold[node]

            if name == disease_input:
                val = 1
            else:
                val = 0
            if val <= threshold:
                recurse(tree_.children_left[node], depth + 1)
            else:
                symptoms_present.append(name)
                recurse(tree_.children_right[node], depth + 1)
        else:
            present_disease = print_disease(tree_.value[node])
            # print( "You may have " + present_disease )
            red_cols = reduced_data.columns
            symptoms_given = red_cols[reduced_data.loc[present_disease].values[0].nonzero()]

```

```

# dis_list=list(symptoms_present)
# if len(dis_list)!=0:
#     print("symptoms present " + str(list(symptoms_present)))
# print("symptoms given " + str(list(symptoms_given)) )
print("Are you experiencing any ")
symptoms_exp=[]
for syms in list(symptoms_given):
    inp=""
    print(syms,"? : ",end="")
    while True:
        inp=input("")
        if(inp=="yes" or inp=="no"):
            break
        else:
            print("provide proper answers i.e. (yes/no) : ",end="")
    if(inp=="yes"):
        symptoms_exp.append(syms)

second_prediction=sec_predict(symptoms_exp)
# print(second_prediction)
calc_condition(symptoms_exp,num_days)
if(present_disease[0]==second_prediction[0]):
    print("You may have ", present_disease[0])
    print(description_list[present_disease[0]])

    # readn(f"You may have {present_disease[0]}")
    # readn(f"{description_list[present_disease[0]]}")

else:
    print("You may have ", present_disease[0], "or ", second_prediction[0])
    print(description_list[present_disease[0]])
    print(description_list[second_prediction[0]])

# print(description_list[present_disease[0]])
precaution_list=precautionDictionary[present_disease[0]]
print("Take following measures : ")
for i,j in enumerate(precaution_list):
    print(i+1,")",j)

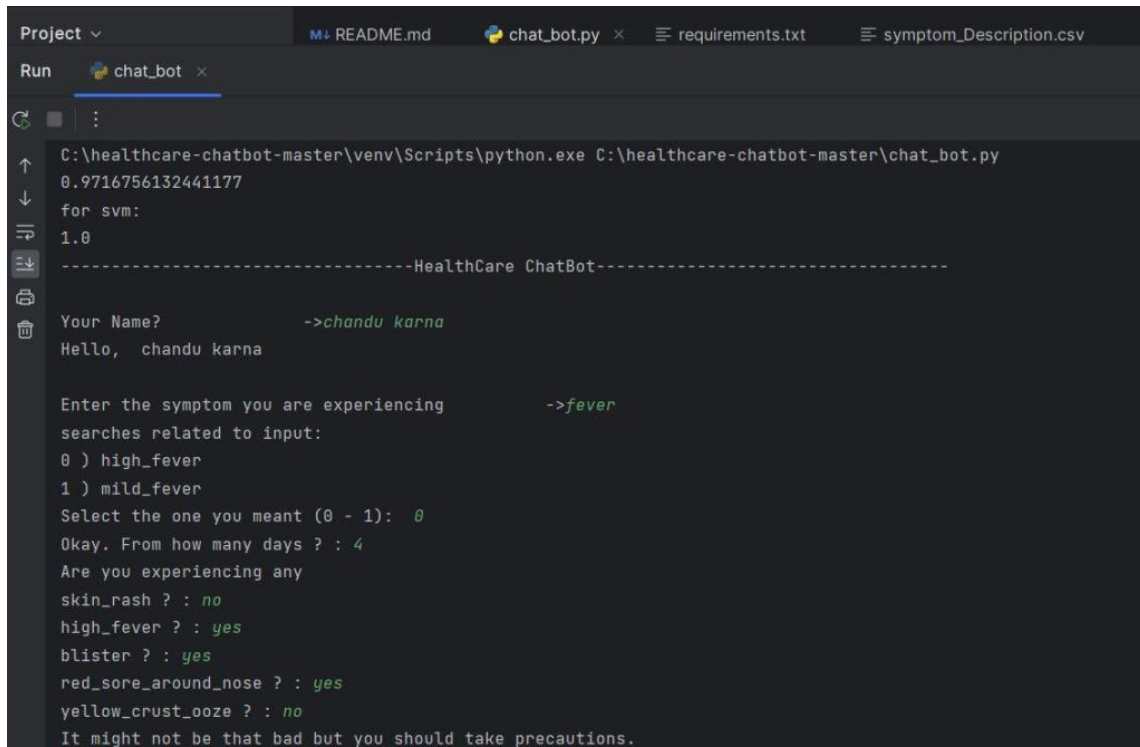
# confidence_level = (1.0*len(symptoms_present))/len(symptoms_given)
# print("confidence level is " + str(confidence_level))

recurse(0, 1)
getSeverityDict()
getDescription()
getprecautionDict()
getInfo()
tree_to_code(clf,cols)
print("-----")

```

## CHAPTER 6

### SCREENSHOTS AND RESULTS:



```
Project v
chat_bot.py x requirements.txt symptom_Description.csv
Run chat_bot x
C:\healthcare-chatbot-master\venv\Scripts\python.exe C:\healthcare-chatbot-master\chat_bot.py
0.9716756132441177
for svm:
1.0
-----HealthCare ChatBot-----
Your Name? ->chandu karna
Hello, chandu karna

Enter the symptom you are experiencing ->fever
searches related to input:
0 ) high_fever
1 ) mild_fever
Select the one you meant (0 - 1): 0
Okay. From how many days ? : 4
Are you experiencing any
skin_rash ? : no
high_fever ? : yes
blister ? : yes
red_sore_around_nose ? : yes
yellow_crust_ooze ? : no
It might not be that bad but you should take precautions.
```

## **CHAPTER 7**

### **CONCLUSION AND FUTURE ENHANCEMENTS**

In conclusion,

The suggested solution helps hospitals and medical centres by enabling patients to ask questions about their health freely via voice or text requests.

The machine gathers data from medical API diagnostic and speaks out with disease treatments. Comparatively speaking, SVM classification accuracy is higher than KNN and Naive Bayes algorithms.

In comparison to other machine learning algorithms, SVM generates 92.33% accuracy and is thus utilised to accurately predict the disease. It also saves time and space.

#### **Future Improvement**

Medical chatbots can be expanded and utilised extensively with other medical systems that allow for prediction by utilising the benefits of the SVM algorithm. It can also be used to plan doctor visits and notify patients of upcoming appointments or checkups.

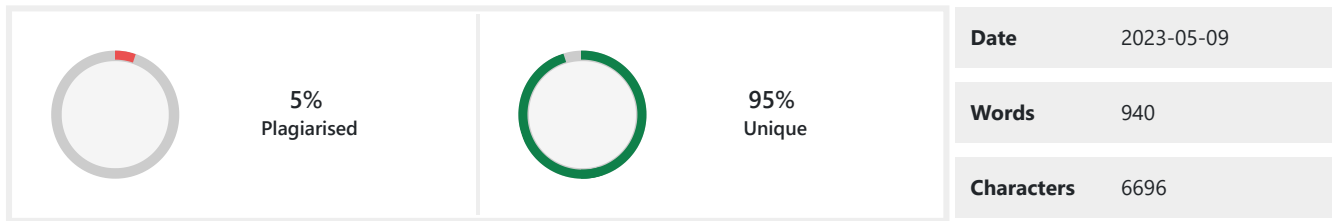
it can also be extended to collect patients' feedback and this will help medical organizations to improve their processes.



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## PLAGIARISM SCAN REPORT



## Content Checked For Plagiarism

A computer programme created for the healthcare sector's chat bot initiative simulates talks with real consumers. The chat bot intends to help patients and healthcare professionals by giving medical advice, responding to inquiries about symptoms and treatments, directing users to resources for healthcare, and helping with appointment scheduling. The need for rapid and easy access to medical information as well as the rising demand for healthcare services have increased the significance of the health chat bot project. Users can communicate with the healthcare system through chat bots, which provide an easy-to-use and accessible method of communication. The health care chat bot project makes use of a number of technologies, including artificial intelligence, machine learning, and natural language processing, to comprehend and reply to user requests. By analysing user inputs and delivering pertinent information, the chatbot can provide precise and prompt responses. Overall, the health care chat bot project is a useful tool for the healthcare sector, offering a low-cost and effective way to provide consumers and healthcare providers with medical advice and support.

### CHAPTER 2 LITERATURE SURVEY

In recent years, the use of chatbots in the healthcare industry has increased. A number of research have been carried out to determine how useful chatbots are in various healthcare settings.

Title of Paper 1: Improving LMS Experience with AIML Base and Retrieval Base R-based chatbot Language

Brief Description: The usage of several techniques, including N-gram, Stemming, TF-IDF, and cosine similarity, is covered in detail in the study. This paper explains how to apply these techniques to produce an optimised outcome rapidly. It details various inquiries and how a chatbot might respond to them. For convenience, they also specified the functional architecture of databases and offered test cases for algorithms. The essence of chatbot quality

Title of Paper 2: Artificial Intelligence Based Personal Assistant,

Brief Description:

Here In this work, a chatbot that acts as a user's personal healthcare assistant is built for healthcare-related purposes. A user can interact with a medical assistant via a dialogue interface. It offers features like ailment diagnosis based on symptoms reported by the user, definitions of medical terms, advice from doctors, scheduling of treatments, and tracking and monitoring of the user's health metrics. Both offline and online performance were looked at overall. To assess the effectiveness and quality of the system, they also ran a number of tests.

To create an eHealth environment for patients' convenience, numerous studies on this topic have been done in the recent years.

Title of Paper 3: Human-to-Machine Conversation Modelling

Brief Description: In this study, hardware and software were combined to develop a chatbot system. This chatbot is Bluetooth-enabled and can move. They controlled chatbots through voice communication. A database that is kept on the Raspberry Pi will be searched using the user's query. This chatbot only functions with particular illnesses like the common cold, typhoid, malaria, etc.

### CHAPTER 3 SYSTEM ARCHITECTURE AND DESIGN

This is the proposed architecture for our AI-Powered Health Chatbot, and we will give further explain for every component in the next upcoming papers. **First of all, the end user interacts with the chatbot through a client**

platform. It's important to be user friendly and give an excellent UX.

Each time, the user is having a request, it's routed to the NLP Engine using the appropriate API's, in the NLP Engine, and with NLU, the chatbot understands the request and format the data into understandable form that can be understood by the Core Engine. Once the Core Engine receives the formatted data, it searches using Deep Learning Algorithms, the appropriate response and send it back to the NLP Engine.

#### CHAPTER 4 METHODOLOGY

The following steps could be used in the technique for creating a health care chatbot:

A health care chatbot's target audience must be determined before any other steps can be taken to develop it.

The chatbot might be created to meet the needs of patients, physicians, nurses, or other healthcare providers.

Designing the chatbot's functionalities would be aided by an understanding of the needs of the target audience.

Identify the chatbot's objectives: The next stage is to identify the objectives of the chatbot.

The objectives should match the requirements of the target audience. For instance, a chatbot created just for patients would be intended to answer general health-related concerns, arrange appointments, and send out reminders.

Gather information and create a database: A substantial amount of data that can aid the chatbot in comprehending and responding to user inquiries is necessary in order to construct one. The information might be found in patient records, clinical data, and medical literature.

By using this information to teach the chatbot, a knowledge base can be created.

Select the chatbot development platform and tools: There are a variety of chatbot development platforms and tools available. The objectives and needs of the chatbot will determine the platform and technologies to use.

Create the dialogue flow for the chatbot: This will establish how the chatbot will communicate with the user. The discussion flow has to be planned to guarantee that the user's inquiries are answered truthfully and effectively.

Create the natural language processing (NLP) system for the chatbot: The chatbot's capacity to comprehend natural language is a crucial feature. The NLP system needs to be able to handle many dialects, accents, and languages.

Test and improve the chatbot: After the chatbot is created, it should be put to the test to make sure it works as intended. Any problems or flaws with the chatbot can be found by conducting user testing. The chatbot can be enhanced and developed based on the input.

Deploy the chatbot: After it has been tested and improved, the chatbot can be made available to the intended audience. To make sure the chatbot is operating properly and answering users' questions, regular checks should be made.

#### CHAPTER 5 CODING AND TESTING

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