

TODEVELOPACHATBOTS USINGPYTHONBASED **ONMACHINELEARNING**

A Report for the Review ETE of Chat Bot making.

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Fall 2021 – 2022**

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ABSTRACT

A **CHATBOT** is a software application used to conduct an on-line chat conversion via text or text-to-speech, in lieu of providing direct contact with a live human agent. This software is used to perform tasks such as quickly responding to users, informing them, helping to purchase products and providing better service to customers. Chatbots, or conversational interfaces as they are also known, present a new way for individuals to interact with computer systems. Traditionally, to get a question answered by a software program involved using a search engine, or filling out a form. A chatbot allows a user to simply ask questions in the same manner that they would address a human. However, chatbots are currently being adopted at a high rate on computer chat platforms.

A chatbot can be used anywhere a human is interacting with a computer system. These are the areas where the fastest adoption is occurring :- human resource, marketing, etc.

We have chosen chatbot as our project because it plays a vital role in our day -day life. And there is a really vast scope of improvement in this field.

As

Chatbots are increasingly present in businesses and often are used to automate tasks that do not require skill-based talents. With customer service taking place via messaging apps as well as phone calls, there are growing numbers of use-cases where chatbot deployment gives organisations a clear return on investment. Call centre workers may be particularly at risk from AI-driven chatbots.

LITERATURE REVIEWS

Chatbots are an AI-powered software program to facilitate conversation with our customers. We can program our chatbots with specific answers to frequently asked questions. The best of chatbots is that we should design this in machine learning – so the customers can get their every type of answer.

In the marketing fields, chatbots are playing a vital role and help very much in business. Chatbots have the great for managing the initial steps of the marketing process, gathering contact details and information for sales calls, answering general customer service questions or providing direction on common tech issues. There can be a huge disconnect between marketing and sales; maybe marketing does not know what sales need to be successful or sales does not know exactly how the marketing funnel is set up (and where prospects are in the sales cycle when they hit their desk).

So to overcome this gap between marketing and sales, chatbots are very best. Especially like in the running Digital era period where it is happening digitally in this digital era. Bots are a more efficient way of gathering information, qualifying leads and setting the sales team up for success. By looping strategy, we can get marketing and sales on the same page and convert more prospects into customers.

So we can conclude that by using chatbots in our digital marketing, it gets more relevant for customers and for newly businesses to set up their business more faster in this competing world.

At last, the motive of the making chatbots to get the success in the digital marketing as the digital marketing is vital nowadays.

PROBLEMS FORMULATION

As we all aware about the fact that, A CHATBOT is a software application used to conduct an on-line chat conversion via text or text-to-speech, in lieu of providing direct contact with a live human agent. These software are used to perform tasks such as quickly responding to users, informing them, helping to purchase products and providing better service to customers.

After spending hour's on this project, we all come to conclusion that Chabot is really cost effective and it is something which is available 24/7 for any clients.

And it can handle multiple client at the same time but there is a problem which we spot that, Sometimes it reply out of the context of the conversation.

so by keeping this in mind we are going to improve it further so that it can reply in precise and concise manner. Which really going to save a lot of time of our clients who are interacting with our Chabot.

And so for that we are going to use python language which helps ultimately to our Chabot to learn independently and it will be easy to update it time to time.

TOOLS FOR IMPLEMENTATION

BASIC REQUIREMENT

To build a chatbot in Python, we have to import all the necessary packages and initialize the variables we want to use in our chatbot project. Also, we are going to work on text data for that, we need to perform data preprocessing on our dataset before designing an ML model.

for that we are going to use tokenizing That helps with text data – it helps fragment the large text dataset into smaller, readable chunks (like words). Once that is done, we can also go for lemmatization that transforms a word into its lemma form. Then it creates a pickle file to store the python objects that are used for predicting the responses of the bot.

STEPS FOR CREATING A CHATBOT:

STEP-1

The first step in creating a chatbot in Python with the ChatterBot library is to install the library in our system.

We are good to install ChatterBot's latest development version directly from GitHub. For this, you will have to write and execute the following command:

```
pip
```

```
install git+git://github.com/gunthercox/ChatterBot.git@master
```

STEP-2

Importing classes is the second step in the Python chatbot creation process.

All we need to do is import two classes – ChatBot

from chatterbot and ListTrainer from chatterbot.trainers.

To do this, we have to execute the following command: `From chatterbot import ChatBot.`

`From chatterbot.trainers import ListTrainer.`

STEP-3

The 3rd step is to Create and Train the Chatbot.

The chatbot we are creating will be an instance of the class “ChatBot.” After creating a new ChatterBot instance, we are going to train the bot to improve its performance.

Training ensures that the bot has enough knowledge to get started with specific responses to specific inputs. We have to execute the following command now:

```
My_bot=ChatBot(name='PyBot',
read_only=True,Logic_adapters=
['chatterbot.logic.MathematicalEvaluation','Chatterbot.logic.
BestMatch'])
```

The command “logic_adapters” denotes the list of adapters used to train the chatbot.

While the “chatterbot.logic.MathematicalEvaluation”

helps the bot to solve math problems, the

“chatterbot.logic.BestMatch” helps it to choose the best match from the list of responses already provided.

STEP-4

The 4th step is to communicate with the Python Chatbot. To interact with Python chatbot, we are going to use the `get_response()` function.

This is how it should look while communicating:

```
Print(my_bot.get_response("hii"))how do you do?  
Print(my_bot.get_response("I feel awesome today"))  
excellent, glad to hear that.
```

However, it is essential to understand that the our chatbot might not know how to answer all our questions.

Since

its knowledge and training is still very limited, so we have to give it time and provide more training data to train it further.

STEP-5

The 5th and last stage is to train our Python Chatbot with a Corpus of Data

for training our python chatbot even further, We are going to use an existing corpus of data. Here's an example of how we train our Python chatbot with a corpus of data provided by the bot itself:

```
From ChatterBot.trainers import ChatterBotCorpusTrainer  
Corpus_trainer = ChatterBotCorpusTrainer(my_bot)  
Corpus_trainer.train('ChatterBot.corpus.english')
```

The good thing is that ChatterBot offers this functionality in many different languages. So, we can also specify a subset of a corpus in some other language also.

MERIT OF PROPOSED SYSTEM

There are many benefits of chatbots. It is used for various purposes in various fields. It is now widely used.

Some of the common uses of chatbots are mentioned below:

Reduced costs:

Chatbots eliminate the need for various employees during online interaction with customers. This obviously has a great impact on the economy of the companies.

24/7 Availability:

Unlike humans, once we install a chatbot, it can handle queries at any time of day.

Learning and updating:

These chatbots are able to learn from interactions and update independently. This is one of the main advantages.

Management of multiple clients:

Humans can serve a limited number of customers at the same time. This restriction does not exist for chatbots, and they can manage all the necessary queries simultaneously.

The purpose of my chatbot is that how people get success in digital marketing. As digital marketing is very vital now days.

As above mentioned benefits we saw that it is more relevant to the user.

Implementation and Description of Project Modules

To implement the project at first we have to install the Chatbot library, then we are gonna follow these steps:-

First, make a file named `train_chatbot.py`. We import the necessary packages for our chatbot and initialize the variables we will use in our Python project.

The data file is in JSON format so we used the `json` package to parse the JSON file into [Python](#).

```
import nltk
from nltk.stem import WordNetLemmatizer
lemmatizer = WordNetLemmatizer()
import json
import pickle
import numpy as np
from keras.models import Sequential
from keras.layers import Dense, Activation, Dropout
from keras.optimizers import SGD
import random
words = []
classes = []
documents = []
ignore_words = ['?', '!']
data_file =
```

This shows our `intents.json` file looks like.

intents.json - D:\dataflair projects\final chatbot\intents.json (3.6.0)

File Edit Format Run Options Window Help

```
{ "intents": [
  { "tag": "greeting",
    "patterns": ["Hi there", "How are you", " s
    "responses": ["Hello, thanks for asking", "G
    "context": [""]
  },
  { "tag": "goodbye",
    "patterns": ["Bye", "See you later", "Goodbye
    "responses": ["See you!", "Have a nice da
    "context": [""]
  },
  { "tag": "thanks",
    "patterns": ["Thanks", "Thank you", "That s
    "responses": ["Happy to help!", "Any time
    "context": [""]
  },
  { "tag": "noanswer",
    "patterns": [],
    "responses": ["Sorry, can't understand yo
    "context": [""]
  },
  { "tag": "options",
    "patterns": ["How you could help me?", "What
    "responses": ["I can guide you through Adver
    "context": [""]
  },
  { "tag": "adverse_drug",
    "patterns": ["How to check Adverse drug reac
    "responses": ["Navigating to Adverse drug re
```

2.Preprocessdata

When working with text data, we need to perform various preprocessing on the data before we make a machine learning or a deep learning model. Tokenizing is the most basic and first thing you can do on text data. Tokenizing is the process of breaking the whole text into small parts like words.

Here we iterate through the patterns and tokenize the sentence using `nlTK.word_tokenize()` function and append each word in the words list. We also create a list of classes for our tags.

```

for intent in
intents['intents']:for pattern in
intent['patterns']:#tokenizeeach
hword
w =
nltk.word_tokenize(pattern)words
ds.extend(w)
#add documents in the
corpusdocuments.append((w,
intent['tag']))#addtoourclasseslist

```

Now we will lemmatize each word and remove duplicate words from the list. Lemmatizing is the process of converting a word into its lemma form and then creating a pickle file to store the Python objects which we will use while predicting.

```

#lemmatize, lowercase each word and remove duplicates
words = [lemmatizer.lemmatize(w.lower()) for w in words if w not in
ignore_words]words= sorted(list(set(words)))
# sort classes
classes=sorted(list(set(classes)))
# documents = combination between patterns and
intentsprint(len(documents),"documents")
# classes= intents
print (len(classes), "classes",
classes)# words
=allwords,vocabulary
print (len(words), "unique lemmatized words",
words)pickle.dump(words,open('words.pkl','wb'))

```

3. Create training and testing data

Now, we will create the training data in which we will provide the input and the output. Our input will be the pattern and output will be the class our input pattern belongs to. But the computer doesn't understand text so we will convert text into numbers.

```

#create our training data

```

```

training = []
# create an empty array for our output
output_empty = [0] * len(classes)
# training set, bag of words for each sentence
for doc in documents:
    # initialize our bag of words
    words_bag = []
    # list of tokenized words for the pattern
    pattern_words = doc[0]
    # lemmatize each word - create base word, in attempt to represent related words
    pattern_words = [lemmatizer.lemmatize(word.lower()) for word in pattern_words]
    # create our bag of words array with 1, if word match found in current pattern
    for w in words:
        bag.append(1) if w in pattern_words else bag.append(0)
    # output is a '0' for each tag and '1' for current tag (for each pattern)
    output_row = list(output_empty)
    output_row[classes.index(doc[1])] = 1
    training.append([bag, output_row])
# shuffle our features and turn into np.array
random.shuffle(training)
training = np.array(training)
# create train and test lists. X - patterns, Y - intents
train_x = list(training[:, 0])
train_y = list(training[:, 1])
print("Training data created")

```

4. Build the model

We have our training data ready, now we will build a deep neural network that has 3 layers. We use the Keras sequential API for this. After training the model for 200 epochs, we achieved 100% accuracy on our model. Let us save the model as 'chatbot_model.h5'.

```

# Create model - 3 layers. First layer 128 neurons, second layer 64 neurons and 3rd output layer contains number of neurons
# equal to number of intents to predict output intent with softmax

```

```

model=Sequential()
model.add(Dense(128, input_shape=(len(train_x[0]),),
activation='relu'))model.add(Dropout(0.5))
model.add(Dense(64,
activation='relu'))model.add(Dropout(0.5))model.add(Dense
nse(len(train_y[0]),activation='softmax'))
# Compile model. Stochastic gradient descent with Nesterov accelerated gradient gives good results
forthismodel
sgd = SGD(lr=0.01, decay=1e-6, momentum=0.9,
nesterov=True)model.compile(loss='categorical_crossentropy', optimizer=sgd,
metrics=['accuracy'])#fitting andsavingthemodel
hist=model.fit(np.array(train_x),np.array(train_y),epochs=200,batch_size=5,verbose=1)model.save('chatbot_m
odel.h5',hist)
print("model created")

```

6.Runthechatbot

To run the chatbot, we have two main files; **train_chatbot.py** and **chatapp.py**.First,wetrainthemodelusingthecommandin theterminal:

```
pythontrain_chatbot.py
```

Ifwedon'tseeanyerrorduringtraining,wehavesuccessfullycreatedthemodel.Thento run theapp,werunthesecondfile.

```
pythonchatgui.py
```

Theprogramwill open up a GUIwindowwithin a fewseconds.WiththeGUIyoucaneasily chatwiththebot.

Screenshots:

Selec C:\Windows\System32\cmd.exe

pythontrain_chatbot.py

D:\dataflair\projects\final_chatbot>

Using TensorFlow backend.

47 documents

9 classes ['adverse_drug', 'blood_pressure', 'blood_pressure_search', 'goodbye', 'hospital_search', 'options', 'pharmacy_search', 'thanks']

88 unique lemmatized words [",", "'", 'a', 'adverse', 'all', 'anyone', 'are', 'e', 'behavior', 'blood', 'by', 'bye', 'can', 'causing', 'chatting', 'check', 'day', 'detail', 'do', 'dont', 'drug', 'entry', 'find', 'for', 'give', 'good', 'have', 'hello', 'help', 'helpful', 'helping', 'hey', 'hi', 'history', 'hola', 'how', 'i', 'id', 'is', 'later', 'list', 'load', 'locate', 'log', 'looking', 'ment', 'me', 'module', 'nearby', 'next', 'nice', 'of', 'offered', 'open', 'pharmacy', 'pressure', 'provide', 'reaction', 'related', 'result', 'search', 'search', 'show', 'suitable', 'support', 'task', 'thank', 'thanks', 'that', 'there', 'to', 'to', 'transfer', 'up', 'want', 'what', 'which', 'with', 'you']

Training data created

2019-11-28 14:10:10.207987: I tensorflow/core/platform/cpu_feature_guard.cc:103] Your TensorFlow binary was not compiled to use AVX2 instructions that this TensorFlow binary was not compiled to use: AVX2

Epoch 1/200

47/47 [=====] - 0s 2ms/step - loss: 2.2080 - accuracy: 0.0000

Epoch 2/200

47/47 [=====] - 0s 211us/step - loss: 2.1478 - accuracy: 0.0000

Epoch 3/200

47/47 [=====] - 0s 228us/step - loss: 2.1427 - accuracy: 0.0000

Hello

— □ ×

You: Hello there. How are

you?Bot: Hi there, how can I

help?You:whatcanyou do?

Bot: I can guide you through Adverse drug
reaction list, Blood pressure tracking,
Hospitals and Pharmacies

You:thanks

youBot:Mypleas

ure

You: see ya got to

go!Bot:See you!


```

PYTHONCODE
import re
import long_response as long

def message_probability(user_message, recognised_words, single_response=False, required_words=[]):
    message_certainty = 0
    has_required_words = True

    # Count how many words are present in each predefined message for word
    in user_message:
        if word in recognised_words: message_certainty += 1

    # Calculates the percent of recognised words in a user message
    percentage = float(message_certainty) / float(len(recognised_words))

    # Check that the required words are in the string for word in required_words:
    if word not in user_message: has_required_words = False; break

    # Must either have the required words, or be a single response if has_required_words or single_response:
    if has_required_words or single_response:
        return int(percentage * 100)
    else:
        return 0

def check_all_messages(message):
    highest_prob_list = {}

    # Simplifies response creation / adds it to the dict
    def response(bot_response, list_of_words, single_response=False, required_words=[]):
        nonlocal highest_prob_list
        highest_prob_list[bot_response] = message_probability(message, list_of_words, single_response, required_words)

    # Responses -----
    response('Hello!', ['hello', 'hi', 'hey', 'sup', 'heyo'], single_response=True)
    response('See you!', ['bye', 'goodbye'], single_response=True)
    response('I\'m doing fine, and you?', ['how', 'are', 'you', 'doing'], required_words=['how'])
    response('You\'re welcome!', ['thank', 'thanks'], single_response=True)
    response('What an exciting personality!, I love him, so much.', ['do', 'u', 'love', 'samir'], required_words=['love'])

    response('yes, he is your project guide', ['shubham'], required_words=['shubham'])
    response('yes, 543 in c block', ['cabin'], required_words=['cabin'])
    response('yetowopunya aatmahjo, aapsabse adhik padhti hbuta apse adhik confused h', ['shruti'], required_words=['shruti'])
    response('yes, he is your reviewer', ['p'], required_words=['p'])

    response(long.R_ADVICE, ['give', 'advice'], required_words=['advice'])
    response(long.R_EATING, ['what', 'you', 'eat'], required_words=['you', 'eat'])

    best_match = max(highest_prob_list, key=highest_prob_list.get)
    print(highest_prob_list)
    # print(f'Best match={best_match} | Score: {highest_prob_list[best_match]}')
    return long.unknown() if highest_prob_list[best_match] < 1 else best_match

# Used to get the response
def get_response(user_input):
    split_message = re.split(r'\s+|[,;?!.-]\s*', user_input.lower())
    response = check_all_messages(split_message)
    return response

# Testing the responses system
while True:

```

```
print('Bot:'+get_response(input('You:')))
```

