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**U**NIVERSITY

Academic Task-3  
(Artificial Intelligence)  
School of Computer Science and Engineering Faculty of  
Technology & Sciences

**Name of the faculty member:-** Mrs. Ankita Wadhawan

**Course Code:-** INT 404 Course

**Title:-** Artificial Intelligence

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**Roll Number:-** 53, 54, 55, 64

**Section:-** K18UW

**GitHub Link:-** <https://github.com/Kprashar/genuine>

## **Abstract:-**

A chatbot is a software application used to conduct an on-line chat conversation via text or text-to-speech, in lieu of providing direct contact with a live human agent.

## **Introduction:-**

### **What is Chatbot?**

A chatbot is an intelligent piece of software that is capable of communicating and performing actions similar to a human. Chatbots are used a lot in customer interaction, marketing on social network sites and instantly messaging the client. There are two basic types of chatbot models based on how they are built; Retrieval based and Generative based models.

#### **1. Retrieval based Chatbots**

A retrieval-based chatbot uses predefined input patterns and responses. It then uses some type of heuristic approach to select the appropriate response. It is widely used in the industry to make goal-oriented chatbots where we can customize the tone and flow of the chatbot to drive our customers with the best experience.

#### **2. Generative based Chatbots**

Generative models are not based on some predefined responses.

They are based on seq to seq neural networks. It is the same idea as machine translation. In machine translation, we translate the source code from one language to another language but here, we are going to transform input into an output. It needs a large amount of data and it is based on Deep Neural networks.

## Literature Review:-

### **What is Natural Language Processing?**

Natural language processing (NLP) is a field of artificial intelligence in which computers analysis, understand, and derive meaning from human language in a smart and useful way. By utilizing NLP, developers can organize and structure knowledge to perform tasks such as automatic summarization, translation, named entity recognition, relationship extraction, sentiment analysis, speech recognition, and topic segmentation.

What is Natural Language Processing good for?

- Summarize blocks of text using Summarizer to extract the most important and central ideas while ignoring irrelevant information.
- Create a chat bot using Parsey McParseface, a language parsing deep learning model made by Google that uses Point-of-Speech tagging.
- **Automatically generate keyword tags** from content using AutoTag, which leverages LDA, a technique that discovers topics contained within a body of text.
- **Identify the type of entity extracted**, such as it being a person, place, or organization using Named Entity Recognition.
- Use Sentiment Analysis to **identify the sentiment of a string of text**, from very negative to neutral to very positive.
- **Reduce words to their root**, or stem, using Porter Stemmer, or **break up text into tokens** using Tokenizer.

### **What is Machine Learning?**

**Machine Learning** is the field of study that gives computers the capability to learn without being explicitly programmed. ML is one of the most exciting technologies that one would have ever come across. As it is evident from the name, it gives the computer that makes it more similar to humans: *The ability to learn*. Machine learning is actively being used today, perhaps in many more places than one would expect.

## **Project Description:-**

In this Python project with source code, we are going to build a chatbot using deep learning techniques. The chatbot will be trained on the dataset which contains categories (intents), pattern and responses. We use a special recurrent neural network (LSTM) to classify which category the user's message belongs to and then we will give a random response from the list of responses.

### **The Dataset**

The dataset we will be using is 'intents.json'. This is a JSON file that contains the patterns we need to find and the responses we want to return to the user.

- Intents.json – The data file which has predefined patterns and responses.
- train\_chatbot.py – In this Python file, we wrote a script to build the model and train our chatbot.
- Words.pkl – This is a pickle file in which we store the words Python object that contains a list of our vocabulary.
- Classes.pkl – The classes pickle file contains the list of categories.
- Chatbot\_model.h5 – This is the trained model that contains information about the model and has weights of the neurons.
- Chatgui.py – This is the Python script in which we implemented GUI for our chatbot. Users can easily interact with the bot.

#### **1. Import and load the data file**

First, make a file name as train\_chatbot.py. We import the necessary packages for our chatbot and initialize the variables we will use in our Python project.

#### **2. Pre-process data**

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.

#### **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.

#### **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'.

#### 5. Predict the response (Graphical User Interface)

Now to predict the sentences and get a response from the user to let us create a new file 'chatgui.py'.

We will load the trained model and then use a graphical user interface that will predict the response from the bot. The model will only tell us the class it belongs to, so we will implement some functions which will identify the class and then retrieve us a random response from the list of responses.

Again we import the necessary packages and load the 'words.pkl' and 'classes.pkl' pickle files which we have created when we trained our model.

#### 6. Run the chatbot

To run the chatbot, we have two main files; train\_chatbot.py and chatgui.py.

### **Conclusion:-**

In this Python data science project, we understood about chatbots and implemented a deep learning version of a chatbot in Python which is accurate. You can customize the data according to business requirements and train the chatbot with great accuracy. Chatbots are used everywhere and all businesses is looking forward to implementing bot in their workflow.

### Output:-

```
Spyder (Python 3.7)
File Edit Search Source Run Debug Consoles Projects Tools View Help

Editor - E:\Kushagr\Semester 4\INT 404 Artificial Intelligence\Project\train_chatbot.py
train_chatbot.py chatgui.py untitled0.py

1 import nltk
2 from nltk.stem import WordNetLemmatizer
3 lemmatizer = WordNetLemmatizer()
4 import json
5 import pickle
6 import tensorflow as tf
7
8 import numpy as np
9 from keras.models import Sequential
10 from keras.layers import Dense, Activation, Dropout
11 from keras.optimizers import SGD
12 import random
13
14 words = []
15 classes = []
16 documents = []
17 ignore_words = ['?', '!', '']
18 data_file = open('intents.json').read()
19 intents = json.loads(data_file)
20
21
22 for intent in intents['intents']:
23     for pattern in intent['patterns']:
24
25         #tokenize each word
26         w = nltk.word_tokenize(pattern)
27         words.extend(w)
28
29         #add documents in the corpus
30         documents.append((w, intent['tag']))
31
32         # add to our classes list
33         if intent['tag'] not in classes:
34             classes.append(intent['tag'])
35
36 # lemmatize and lower each word and remove duplicates
37 words = [lemmatizer.lemmatize(w.lower()) for w in words if w not in ignore_words]
38 words = sorted(list(set(words)))
39 # sort classes
40 classes = sorted(list(set(classes)))
41 print (len(documents), "documents")
42 # classes = intents
```

### Usage

Here you can get help of any object by pressing **Ctrl+H** in front of it, either on the Editor or the Console.

Help can also be shown automatically after writing a left parenthesis next to an object. You can activate this behavior in [Preferences > Help](#).

New to Spyder? Read our [tutorial](#)

Variable explorer | File explorer | Help

Python console

Console 2/A

```
47/47 [=====] - 0s 0us/step - loss: 0.0746 - accuracy: 0.987
Epoch 196/200
47/47 [=====] - 0s 0us/step - loss: 0.0945 - accuracy: 0.9574
Epoch 197/200
47/47 [=====] - 0s 565us/step - loss: 0.0690 - accuracy:
0.9574
Epoch 198/200
47/47 [=====] - 0s 276us/step - loss: 0.0353 - accuracy:
1.0000
Epoch 199/200
47/47 [=====] - 0s 276us/step - loss: 0.0878 - accuracy:
0.9574
Epoch 200/200
47/47 [=====] - 0s 254us/step - loss: 0.0391 - accuracy:
1.0000
model created

In [24]: runfile('E:\Kushagr\Semester 4\INT 404 Artificial Intelligence\Project\
chatgui.py', wdir='E:\Kushagr\Semester 4\INT 404 Artificial Intelligence\Project')

In [25]: runfile('E:\Kushagr\Semester 4\INT 404 Artificial Intelligence\Project\
chatgui.py', wdir='E:\Kushagr\Semester 4\INT 404 Artificial Intelligence\Project')

In [26]:
```

Python console | History log

Dimensions: BW End-of-line: CRLF Encoding: ASCII Line: 35 Column: 55 Memory: 74%

The image shows the Spyder Python IDE interface. The main editor displays a Jupyter Notebook with a Python script for training a neural network. The script includes the following code:

```
39 classes = sorted(list(set(classes)))
40 # documents = combination between patterns and intents
41 print (len(documents), "documents")
42 # classes = intents
43 print (len(classes), "classes", classes)
44 # words = all words, vocabulary
45 print (len(words), "unique lemmatized words", words)
46
47
48 pickle.dump(words,open('words.pkl','wb'))
49 pickle.dump(classes,open('classes.pkl','wb'))
50
51 # create our training data
52 training = []
53 # create an empty array for our output
54 output_empty = [0] * len(classes)
55 # training set, bag of words for each sentence
56 for doc in documents:
57     # initialize our bag of words
58     bag = []
59     # list of tokenized words for the pattern
60     pattern_words = doc[0]
61     # lemmatize each word - create base word, in attempt to represent related words
62     pattern_words = [lemmatizer.lemmatize(word.lower()) for word in pattern_words]
63     # create our bag of words array with 1, if word match found in current pattern
64     for w in words:
65         bag.append(1) if w in pattern_words else bag.append(0)
66
67 # output is a '0' for each tag and '1' for current tag (for each pattern)
68 output_row = list(output_empty)
69 output_row[classes.index(doc[1])] = 1
70
71 training.append([bag, output_row])
72 # shuffle our features and turn into np.array
73 random.shuffle(training)
74 training = np.array(training)
75 # create train and test lists. X - patterns, Y - intents
76 train_x = list(training[:,0])
77 train_y = list(training[:,1])
78 print("Training data created")
79
80
```

The Python console on the right shows the output of the script:

```
In [24]: runfile('E:/Kushagr/Semester 4/INT 404 Artificial Intelligence/Project/
chatgui.py', wdir='E:/Kushagr/Semester 4/INT 404 Artificial Intelligence/Project')

In [25]: runfile('E:/Kushagr/Semester 4/INT 404 Artificial Intelligence/Project/
chatgui.py', wdir='E:/Kushagr/Semester 4/INT 404 Artificial Intelligence/Project')

In [26]:
```

The console output for the training process is as follows:

```
47/47 [=====] - 0s 0us/step - loss: 0.0746 - accuracy: 0.9787
Epoch 196/200
47/47 [=====] - 0s 0us/step - loss: 0.0945 - accuracy: 0.9574
Epoch 197/200
47/47 [=====] - 0s 565us/step - loss: 0.0690 - accuracy:
0.9574
Epoch 198/200
47/47 [=====] - 0s 276us/step - loss: 0.0353 - accuracy:
1.0000
Epoch 199/200
47/47 [=====] - 0s 276us/step - loss: 0.0878 - accuracy:
0.9574
Epoch 200/200
47/47 [=====] - 0s 254us/step - loss: 0.0391 - accuracy:
1.0000
model created
```

Spyder (Python 3.7)

File Edit Search Source Run Debug Consoles Projects Tools View Help

Editor - E:\Kushagr\Semester 4\INT 404 Artificial Intelligence\Project\train\_chatbot.py

```
59 # list of tokenized words for the pattern
60 pattern_words = doc[0]
61 # lemmatize each word - create base word, in attempt to represent related words
62 pattern_words = [lemmatizer.lemmatize(word.lower()) for word in pattern_words]
63 # create our bag of words array with 1, if word match found in current pattern
64 for w in words:
65     bag.append(1) if w in pattern_words else bag.append(0)
66
67 # output is a '0' for each tag and '1' for current tag (for each pattern)
68 output_row = list(output_empty)
69 output_row[classes.index(doc[1])] = 1
70
71 training.append([bag, output_row])
72 # shuffle our features and turn into np.array
73 random.shuffle(training)
74 training = np.array(training)
75 # create train and test lists, X - patterns, Y - intents
76 train_x = list(training[:,0])
77 train_y = list(training[:,1])
78 print("Training data created")
79
80
81 # Create model - 3 layers. First layer 128 neurons, second layer 64 neurons and 3rd output layer contains number
82 # equal to number of intents to predict output intent with softmax
83 model = Sequential()
84 model.add(Dense(128, input_shape=(len(train_x[0]),), activation='relu'))
85 model.add(Dense(64, activation='relu'))
86 model.add(Dense(64, activation='relu'))
87 model.add(Dense(len(train_y[0]), activation='softmax'))
88
89 # Compile model. Stochastic gradient descent with Nesterov accelerated gradient gives good results for this model
90 sgd = SGD(lr=0.01, decay=1e-6, momentum=0.9, nesterov=True)
91 model.compile(loss='categorical_crossentropy', optimizer=sgd, metrics=['accuracy'])
92
93 # fitting and saving the model
94 hist = model.fit(np.array(train_x), np.array(train_y), epochs=200, batch_size=5, verbose=1)
95 model.save('chatbot_model.h5', hist)
96
97 print("model created")
98
```

train\_chatbot.py chatgui.py untitled0.py

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Variable explorer File explorer Help

Python console

Console 2/A

```
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Epoch 198/200
47/47 [=====] - 0s 276us/step - loss: 0.0353 - accuracy: 1.0000
Epoch 199/200
47/47 [=====] - 0s 276us/step - loss: 0.0878 - accuracy: 0.9574
Epoch 200/200
47/47 [=====] - 0s 254us/step - loss: 0.0391 - accuracy: 1.0000
model created

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In [26]:
```

Permissions: RW End-of-lines: CRLF Encoding: ASCII Line: 35 Column: 55 Memory: 73 %

Spyder (Python 3.7)

File Edit Search Source Run Debug Consoles Projects Tools View Help

Editor - E:\Kushagr\Semester 4\INT 404 Artificial Intelligence\Project\chatgui.py

```
1 import nltk
2 from nltk.stem import WordNetLemmatizer
3 lemmatizer = WordNetLemmatizer()
4 import pickle
5 import numpy as np
6
7 from keras.models import load_model
8 model = load_model('chatbot_model.h5')
9 import json
10 import random
11 intents = json.loads(open('intents.json').read())
12 words = pickle.load(open('words.pkl', 'rb'))
13 classes = pickle.load(open('classes.pkl', 'rb'))
14
15
16 def clean_up_sentence(sentence):
17     # tokenize the pattern - split words into array
18     sentence_words = nltk.word_tokenize(sentence)
19     # stem each word - create short form for word
20     sentence_words = [lemmatizer.lemmatize(word.lower()) for word in sentence_words]
21     return sentence_words
22
23 # return bag of words array: 0 or 1 for each word in the bag that exists in the sentence
24
25 def bow(sentence, words, show_details=True):
26     # tokenize the pattern
27     sentence_words = clean_up_sentence(sentence)
28     # bag of words - matrix of N words, vocabulary matrix
29     bag = [0]*len(words)
30     for s in sentence_words:
31         for i,w in enumerate(words):
32             # assign 1 if current word is in the vocabulary position
33             bag[i] = 1
34             if show_details:
35                 print("found in bag: %s" % w)
36     return(np.array(bag))
37
38
39 def predict_class(sentence, model):
40     # filter out predictions below a threshold
41     p = bow(sentence, words, show_details=False)
42     res = model.predict(np.array([p]))[0]
43     ERROR_THRESHOLD = 0.25
```

train\_chatbot.py chatgui.py untitled0.py

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Variable explorer File explorer Help

Python console

Console 2/A

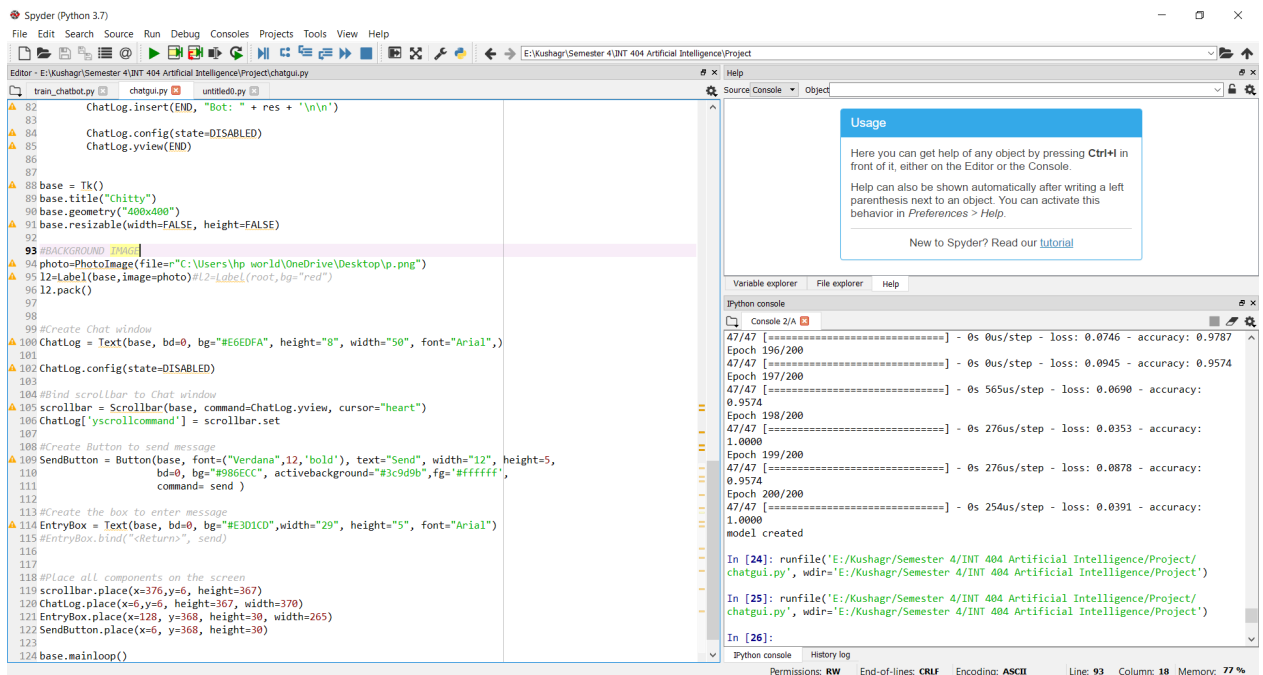
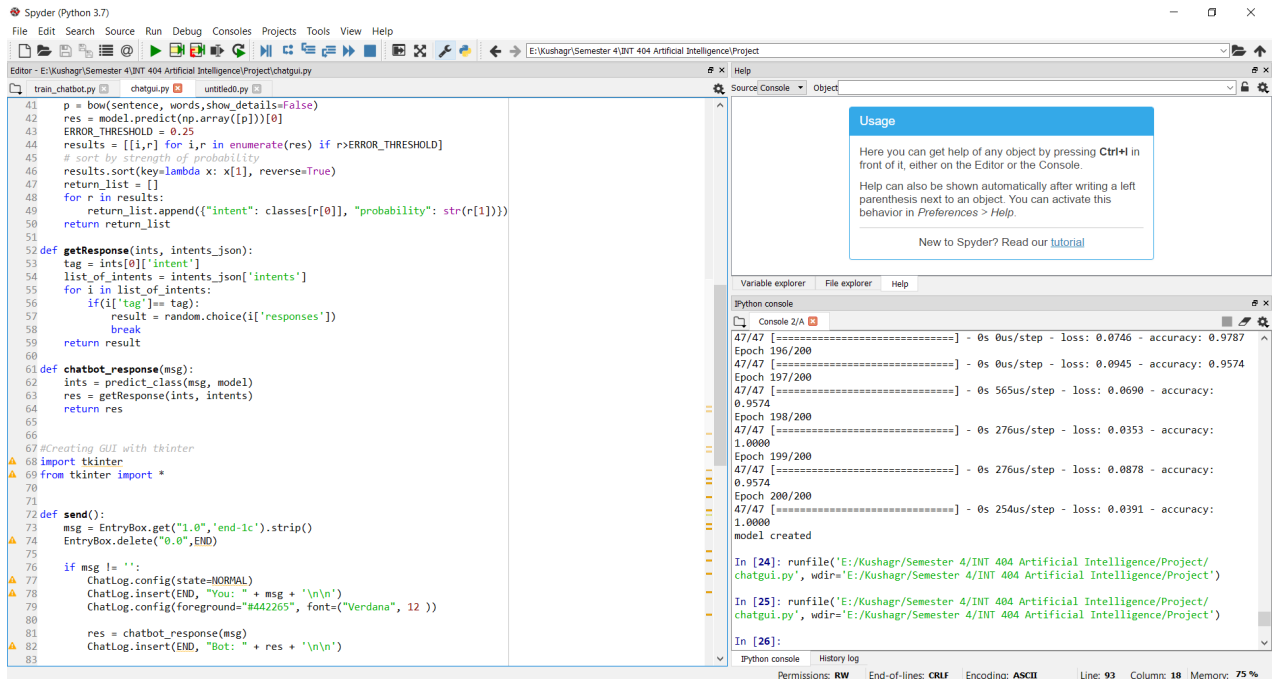
```
47/47 [=====] - 0s 0us/step - loss: 0.0746 - accuracy: 0.9787
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In [26]:
```

Permissions: RW End-of-lines: CRLF Encoding: ASCII Line: 93 Column: 18 Memory: 75 %





Chitty



You: Hi

Bot: Good to see you again

You: What can you do

Bot: I can help you to know some scerect of the life

Send

Tell me one life lesson

Chitty



You: Hi

Bot: Good to see you again

You: What can you do

Bot: I can tell you some life facts

You: Tell me one life lesson

Bot: Make improvements, not excuses. Seek respect, not attention.

You: Tell me one more life lesson

Bot: Life has no remote....get up and change it yourself!

You: Tell me something else

Send



Chitty



Bot: Good to see you again.

You: What can you do

Bot: I can tell you some life facts

You: Tell me one life lesson

Bot: Make improvements, not excuses. Seek respect, not attention.

You: Tell me one more life lesson

Bot: Life has no remote....get up and change it yourself!

You: Tell me something else

Bot: you are time killer

Send