Hamna Mustafa

Hbm170002

CS4395 – Human Language Technologies

November 13th 2022

Chatbot Report

I created a chatbot using the information retrieval method. At a very high level, I created a knowledge base and used NLP techniques to format and retrieve queries. I then connected my chatbot to a channel in an IRC (Internet Relay Chat) server known as Libera Chat where users could log on and converse with it.

Running Instructions:

First run KnowledgeBaseTrain.py using the following command:

python3 KnowledgeBaseTrain.py

Then run main.py:

python3 main.py

This will connect the chatbot to the IRC server. You can then go to this link:

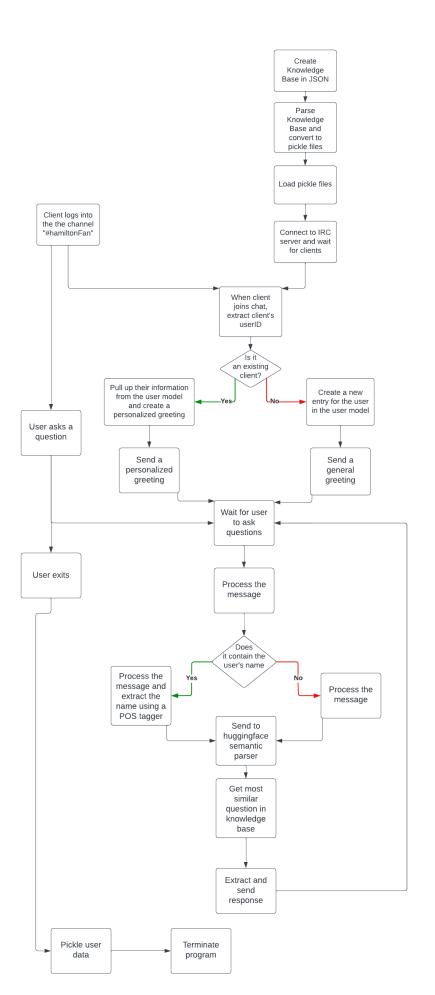
https://web.libera.chat

Then join the channel "#hamiltonFan

Now you can start chatting! Enjoy!

System Description

This project had a number of modules including the knowledge base, information retrieval, processing queries, user models, and an irc server. Below is the logic tree of the program:



Knowledge Base:

Since my chatbot used the information retrieval method, it needed a knowledge base to retrieve information from. I decided to use the same topic as my web crawler project for this chatbot which was about "Hamilton: An American Musical". Hence, my chatbot would be a fanbot for Hamilton, providing the user with information regarding the musical.

In order to make it easy for my chatbot to retrieve information, I decided to create a new knowledge base that was in JSON format. I divided the information up by tags, created some example questions, and wrote out several plausible responses to those questions.

Thus, each entry in the knowledge base had a tag, patterns (example queries), and responses. Here is a screenshot of what the knowledge base looks like:

Once this JSON file was created, I parsed the file and extracted all the patterns along with the corresponding tags and put them into 2 arrays. So, one array had all the queries and the array contained the tags corresponding to those queries. For example:

```
training_sentences = ["hi", "hey", "is anyone there", "who are you", "see you later"] training_label = ["greeting", "greeting", "greeting", "about", "goodbye"]
```

Thus, you can see the the first three phrases in 'training_sentence' had the tag 'greeting' and so on. I then pickled these arrays into 2 separate pickle files. These files became the part of the knowledge base that my chatbot would directly interact with.

Retrieving Information:

Now that I had a knowledge base to work from, I needed to be able to take the messages sent by the user and retrieve the most relevant information from the knowledge base. It took me a while to figure out how to approach this. I first tried a machine learning approach where the tags were the target and the training sentences were the predictors. I tried many different models such as Naïve Bayes, Logistic Regression, and Neural networks but no matter how many models and topologies I tried, I was getting terrible accuracies. Eventually, I realized that there were two major reasons for this: 1) I didn't have enough data in my knowledge base 2) The queries needed some sort of semantic parsing. Once I realized the second reason, I knew I had to find some way to semantically compare the user's message with all the data in the knowledge base to find the question most similar in meaning and retrieve the answer. This led me to Huggingface's "Sentence Similarity". They had an API that would take the user message and the array of training sentences and return an array which contained the semantic similarity values between the message and the training sentences. For example, if the user message was: "That is a happy person" and the training sentences were: ["That is a happy dog, "That is a very happy person", "Today is a sunny day", then the Sentence Similarity API would return: [0.853, 0.981, 0.655]. As you can see, "That is a very happy person" got the highest score which makes sense because it is the most semantically similar to the user message.

Mustafa 5

Using this method proved to be very fruitful. Once I would receive all the semantic similarity values, I would find the max value and the corresponding training sentence. I would then pick a response from the knowledge base of that training sentence.

Processing Queries:

Once I had figured out how to find relevant answers, I had to process the user's message to increase the chances of correct retrievals. I used nltk to do this. The processing that I was most concerned with was when the user message contained the user's name. I will discuss the user model later but all you need to know for now is that I needed to be able to extract the user's name. If the user sent this message: "My name is Hamna", I needed to be able to parse it and find the name. I used two things to do this: nltk's POS tagger and a variable prevTag. PrevTag was a variable that contained the tag of the previous user message. The way that my knowledge base is designed, if the user asks the bot it's name, the bot responds and asks the user for their name.

For example:

User: "What should I call you?"

Bot: "You can call me Tara. What can I call you?"

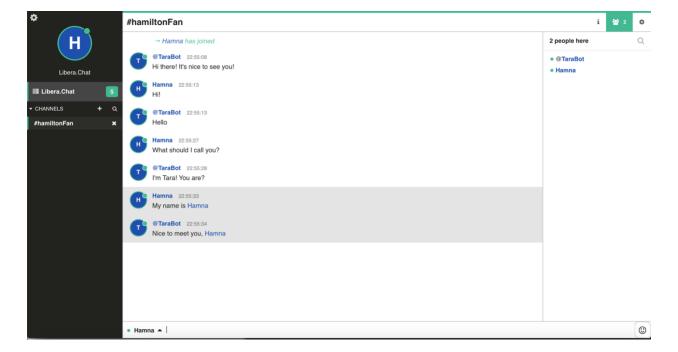
User: "My name is Hamna"

Thus, the first user message fell into the tag "name" and I needed to make sure the second one fell into the tag "theirname". I couldn't directly send the message to the Semantic Similarity parser because every user's name would be different and that would increase the chance for error. Thus, in the training sentences for "theirname", I replaced where the name would be by "nnp". Then, if the prevTag was "name", I used nltk's POS tagging to find the proper noun (NNP) in the user's sentence and replace it with 'nnp'. So if the message was: "My name is Hamna", it would be formatted to become "My name is nnp". I also saved the name so that it

could be used for the user model (which I will talk about in detail later) and so that I could add it to the response string.

As you can see in the "theirname" responses, it leaves space for the user's name to be concatenated. So if the first response was chosen for the conversation above, the message sent to the user would be: "*Nice to meet you, Hamna*".

Finally, I converted all the user messages to lowercase (except for names) before sending them to hugging face.



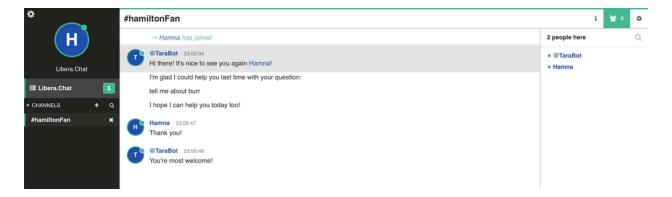
User Model:

As per the project's description, I had to save each user's information which could later be used to personalize the chatbot's responses according to each user. I decided to use a pickled dictionary to do this, with the key being the user's id. When the program starts, it unpickles the "userData.pickle" if it exists which contains information about users from previous runs. This is the format of each user's entry:

```
{
"name": ""

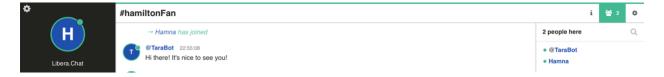
"queries": []
"responses": []
}
```

Everytime a user joins the chat, the program isolates the user's ip address and saves it as userid. It then checks if that userid is already in the user model. If it is, then the bot retrieves that information and sends the user a personalized greeting. It uses the user's name in the greeting and also mentions the last question the user asked.



If it is a new user that does not yet exist in the user model, the bot creates an entry for that user which is then filled as the user asks questions and provides information. For the queries and

responses, I only stored queries relative to Hamilton and not general ones such as "Hello" or "What's your name". For new users, the bot sends a general greeting.



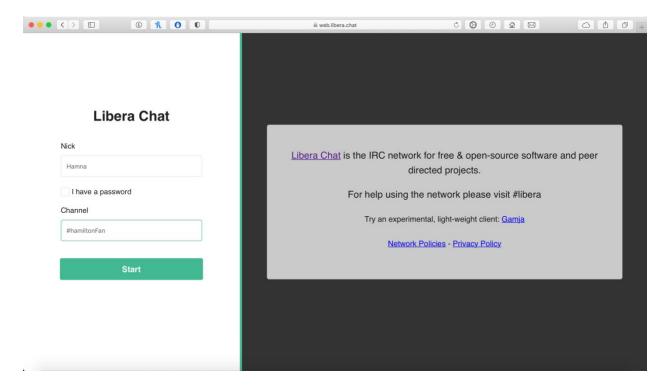
Once the user exits the channel, the bot once again pickles the user model to be used for later runs. It then terminates well.

IRC Server:

Simply writing the bot on my local machine and using the console to converse with it was not enough, I needed to be able to host the bot somewhere so that it could communicate with different users. I decided to connect it to an IRC server using the socket library. The IRC server that I chose was Libera Chat which can be accessed by the following url:

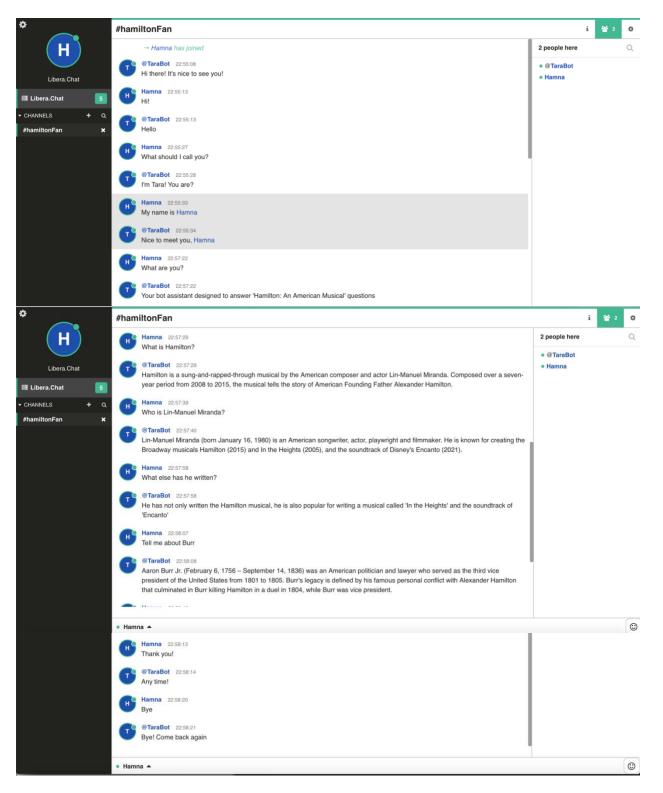
https://web.libera.chat

The channel that I connected to was "#hamiltonFan". Thus, in order to chat with my bot, all the user needs to do is go to the above url and connect to the channel "#hamiltonFan".



The bot connects to this channel when I run the program and stays connected until a user leaves after which it terminates.

Sample Dialog Interaction



User Model after the above interaction

```
{"47-185-246-201.dlls.tx.frontiernet.net": {
       "name": "Hamna",
       "queries": ["what is hamilton?\r\n", "who is lin-manuel
          miranda?\r\n", "what else has he written?\r\n", "
           tell me about burr\r\n"],
      "responses": ["Hamilton is a sung-and-rapped-through
           musical by the American composer and actor
          Lin-Manuel Miranda. Composed over a seven-year
          period from 2008 to 2015, the musical tells the
           story of American Founding Father Alexander
          Hamilton.", "Lin-Manuel Miranda (born January 16,
           1980) is an American songwriter, actor, playwright
           and filmmaker. He is known for creating the
           Broadway musicals Hamilton (2015) and In the
          Heights (2005), and the soundtrack of Disney's
          Encanto (2021).", "He has not only written the
          Hamilton musical, he is also popular for writing a
          musical called 'In the Heights' and the soundtrack
           of 'Encanto'", "Aaron Burr Jr. (February 6, 1756
           \u2013 September 14, 1836) was an American
           politician and lawyer who served as the third vice
           president of the United States from 1801 to 1805.
          Burr's legacy is defined by his famous personal
           conflict with Alexander Hamilton that culminated in
           Burr killing Hamilton in a duel in 1804, while Burr
           was vice president."]}}
5
```

The user model is actually stored as a pickle file. I converted it to JSON in the above screenshot for easy in reading as pickle files are in binary.

Appendix

Knowledge Base

Although I have attached examples of what the knowledge base looks like above, here are a few more.

```
{"tag": "aboutburr",
"patterns": ["who was aaron burr?", "tell me about burr", "i
    want ot know about aaron burr"],
"responses": ["Aaron Burr Jr. (February 6, 1756 - September 14,
    1836) was an American politician and lawyer who served as
                                    the third vice president of the United States from 1801 to
                                   1805. Burr's legacy is defined by his famous personal conflict with Alexander Hamilton that culminated in Burr killing Hamilton in a duel in 1804, while Burr was vice president."]
                      },
{"tag": "whenburr",
"patterns": ["what was aaron burr's age", "I want to know about
   burr's age", "when was burr born", "whats burr's birthday", "
   how old was burr"],
"responses": ["Aaron Burr Jr. was born on February 6, 1756 and
   died on September 14, 1836. He was 80 years old when he died.
   "]
69
70
71
                      },
{"tag": "aboutwashington",
"patterns": ["who was washinton?", "tell me about george
    washington", "i want to know about washington"],
"responses": ["George Washington (February 22, 1732[b] —
    December 14, 1799) was an American military officer,
    statesman, and Founding Father who served as the first
    president of the United States from 1789 to 1797."]
.
73
74
75
                       77
50
51
                       namilton musical"],
"responses": ["Hamilton is a sung-and-rapped-through musical by
the American composer and actor Lin-Manuel Miranda. Composed
over a seven-year period from 2008 to 2015, the musical
tells the story of American Founding Father Alexander
Hamilton."]
                     53
54
55
                      },
{"tag": "whenhamilton",
"patterns": ["when was hamilton written", "what year was
    hamilton released", "when was hamilton released", "when did
    hamilton come out"],
"responses": ["It premiered Off-Broadway on February 17, 2015,
    at the Public Theater in Lower Manhattan"]
57
58
59
```

User Model

Example of User Model

```
{"47-185-246-201.dlls.tx.frontiernet.net": {
        "name": "Hamna",
2
        "queries": ["what is hamilton?\r\n", "who is lin-manuel
            miranda?\r\n", "what else has he written?\r\n", "
             tell me about burr\r\n"],
        "responses": ["Hamilton is a sung-and-rapped-through
            musical by the American composer and actor
Lin-Manuel Miranda. Composed over a seven-year
period from 2008 to 2015, the musical tells the
             story of American Founding Father Alexander
            Hamilton.", "Lin-Manuel Miranda (born January 16,
             1980) is an American songwriter, actor, playwright
             and filmmaker. He is known for creating the
             Broadway musicals Hamilton (2015) and In the
            Heights (2005), and the soundtrack of Disney's
            Encanto (2021).", "He has not only written the Hamilton musical, he is also popular for writing a
            musical called 'In the Heights' and the soundtrack of 'Encanto'", "Aaron Burr Jr. (February 6, 1756
             \u2013 September 14, 1836) was an American
             politician and lawyer who served as the third vice
             president of the United States from 1801 to 1805.
             Burr's legacy is defined by his famous personal
             conflict with Alexander Hamilton that culminated in
             Burr killing Hamilton in a duel in 1804, while Burr
            was vice president."]}}
```

Evaluation

After testing my project, I have made the following evaluation:

Strengths:

- The chatbot replies very quickly fast lookup
- Semantic Similarity works well
- IRC Server provides ease in communication
- Use of pickle files increases efficiency
- Storing user model data makes the chat personable

Weaknesses:

- Limited knowledge base
- User messages that are not similar to knowledge base can provide incorrect responses
- The chatbot needs to be running on my local machine

All in all, I believe that the biggest weakness was the limited knowledge base. If I had a larger, more extensive knowledge base, the chatbot would be able to handle many types of questions and answer correctly. My chatbot did work quite fast on my machine and I believe that semantic similarity parsing was the way to go!