Logos generated by Dall-E

Chatbot for a fictive company specialized in health data

During this project, I wanted to design a chatbot that would answer a user's questions about a fictive company's products. This was a very interesting project, as it covered the development of a product from start to finish, in this case deciding and writing the specs, creating the chatbot, developing the backend and then building a lightweight frontend.

This project has been inspired from codebasics (https://codebasics.io/), who built a chatbot for food delivery.

You can directly access to the chatbot following this link: https://clement1103.github.io/frontend/, but I would strongly advise you to read the whole document.

1. The fictive company

First, it was necessary to define the context in which the chatbot would be employed. It would be used by a fictitive company specializing in healthcare AI, offering three flagship services:

MedIAg

An Al tool that analyzes medical data to assist doctors in making accurate diagnoses while improving over time



SyntheMedix

A tool which generates synthetic healthcare data that mimics real data patterns to protect patient privacy and support research.



ForecastMed

A predictive analytics tool that helps healthcare providers anticipate patient outcomes and optimize care strategies using medical data and machine learning.



These products would be proposed to other companies specialized in AI for health, or medical institutions in order to improve their efficiency.

Also, the company would offer custom services that meet the specific needs of other collaborators.

2. Definition of the specs

It was also necessary to clearly define the requirements that the chatbot would have to meet. It's obvious that I couldn't develop a chatbot that could understand several languages, that would have a very wide range of answers to suit all the questions it was asked.

After analyzing and prioritizing the initial requirements, I decided that the chatbot would have to meet the following criteria:

- Present the different services available,
- Capture the user's interest following each presentation,
- Obtain and confirm user information so that an expert can contact them again,
- Redirect the user to an expert if a complex question is asked
- · Schedule a meeting with an expert.

Here are 3 simplified conversation flows that the chatbot should be able to handle:

BOT: Welcome. Would you like a presentation of the services or to schedule a meeting?

USER: Presentation.

BOT: Here are our products [...] Do you

want to know more about one?

USER: Tell me more about SyntheMedix. **BOT**: [Presentation of the product] Are you

interested? **USER**: Yes.

BOT: Do you want to discover another

service?

USER: Yes, tell me more about

ForecastMed.

BOT: [Presentation of the product] Are you

interested? **USER**: No.

BOT: Do you want to discover another

service? **USER**: No.

BOT: What are your email/phone number?

USER: contact1@example.com,

0123456789

BOT: Are these information correct?

USER: Yes

BOT: Someone will contact you.

BOT: Welcome. Would you like a presentation of the services or to schedule a meeting?

USER: Schedule a meeting.

BOT: What are your email/phone number?

USER: contact1@example.com,

0123456789

BOT: Are these information correct?

USER: Yes.

BOT: Someone will contact you.

BOT: [Presentation of the product] Are you interested?

(3)

USER: How precisely do you think it would

behave in this specific case?

BOT: I don't have the answer. Do you want an expert to get in touch with you to answer this specific question?

USER: Yes please.

BOT: What are your email/phone number?

USER: contact1@example.com,

0123456789

BOT: Are these information correct?

USER: Yes.

BOT: Someone will contact you.

Then, the useful information from the chat (user coordinates, products of interest, date of the chat) would be extracted and host in a SQL database.

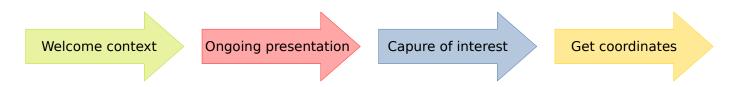
Finally, a lightweight frontend will be developped in order to complete the production launch of the chatbot.

3. Chatbot creation using dialogflow

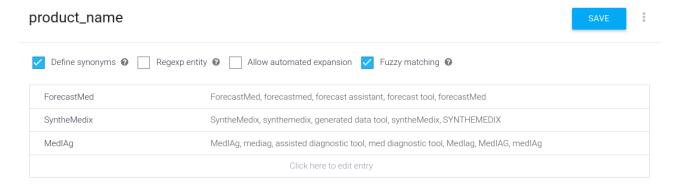
During this project, I will use Dialogflow which is a NLP platform developed by Google, allowing developers to create conversational interfaces such as chatbots and voice assistants. It makes it easy to understand user intentions and manage dialogues, while integrating machine learning features to improve interactions over time.

To understand the development of the chatbot, there are some key concepts that need to be presented.

- Contexts: Used to manage conversation status and help keep track of the information needed to respond appropriately, enabling more natural and continuous dialogues.
 - In our case, here are what the contexts could be like (we'll see later that it has slightly been modified):



• Entities (or NER - Named Entity Recognition): Extract specific information from user input, such as names, dates, places or other relevant terms. For example, in the sentence "Book a table for two people in Paris", "two people" and "Paris" would be entities. One can also define custom entities, specific to the use case. This is what I did to define the services:



For each entity, we define a name, and its corresponding synonyms. It allows the chatbot to identify the different services even if the entity is misspelled, called with another name or with lower/upper case.

forecastmed
ForecastMed
Forecast tool
Forecast assistant

synthemedix
syntheMedix
SyntheMedix
generated data tool
SYNTHEMEDIX

Assisted diagnostic tool

MedIAg

MedIag medIAg

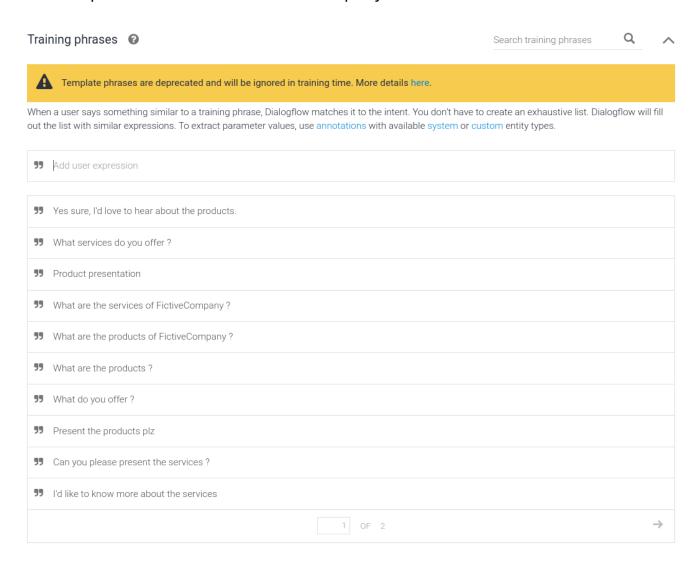
MedIAG

The entity recognition will also be useful when obtaining the email address and the phone number of the user.

 Intents: Represent user intentions. Each intent is associated with example inputs that the user might say, helping Dialogflow to understand what the user wants to do. An intent is defined by training phrases and a response.

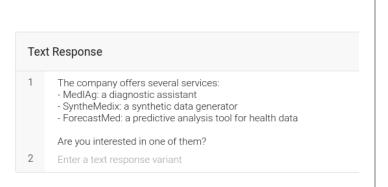
The training phrases are sentences that the user might use, which will trigger the intent. For each intent, it is advisable to define several training phrases, in order to cover the full range of user response. However, the user phrases don't need to be exactly as the training ones, because Dialogflow uses NLP algorithms to detect sentences sharing the same meaning.

Here is an example of training phrases used to trigger the intent aiming to present the services of the company:



With this set of training phrases, if the user type something like « I would like a presentation of the products your offer please. », the intent will be triggered even if this sentence is not in the training set.

Then, for each intent, we associate one response or more.

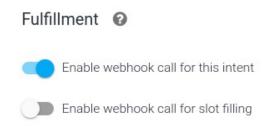


It can be a simple text response...

... or a more complex one, called custom payload, allowing to display different answer styles such as multiple choices.

To make the triggering of the desired intent easier, one can define an input and an output contexts during the creation of the intent.

 Fulfillment: Allows to execute actions outside Dialogflow, such as calling APIs or interacting with databases, to generate dynamic responses based on user input. This will be very useful for gathering and storing user information in our database. To link our chatbot with the backend, it is necessary to activate the fulfillment on the intent definition (but not every intent requires to access to the backend).



 Fallback intents: Triggers when the chatbot can't categorize the user input into any intent.

Finally, here is the list of the intents I defined:

| • | capture.of.disinterest - context: ongoing presentation | | | |
|---|--|--|--|--|
| • | capture.of.interest - context: ongoing presentation | | | |
| • | check.coordinates | | | |
| • | complete.presentation - context: interest-capture | | | |
| • | coordinates.correct | | | |
| • | coordinates.incorrect | | | |
| | Default Fallback Intent | | | |
| • | Default Welcome Intent | | | |
| | fallback Contexts: ongoing-presentation | | | |
| | fallback2 Contexts: interest-capture | | | |
| • | no.appointment - context: fallback-presentation | | | |
| • | presentation.ForecastMed - context: interest-capture | | | |
| • | presentation.ForecastMed - context: ongoing-presentation | | | |
| • | presentation.MedIAg - context: interest-capture | | | |
| • | presentation.MedIAg - context: ongoing-presentation | | | |
| • | presentation.SyntheMedix - context: interest-capture | | | |
| • | presentation.SyntheMedix - context: ongoing-presentation | | | |
| • | product.presentation - choices | | | |
| • | set.appointment | | | |
| • | set.appointment - context: fallback-presentation | | | |

Sometimes, an intent is defined twice, for different contexts.

For instance, the intent « presentation SyntheMedix – context: ongoing-presentation » triggers at the beginning of the chat, when the user wants to know more about SyntheMedix after requesting a presentation of the different products, whereas « presentation SyntheMedix – context: interest-capture » triggers once the presentation of a product has already be done, the user has declared is (dis)interest in the product, and wants to discover another one.

4. Backend development

As mentioned above, the backend's role will be to enter the list of products that the customer is interested in, obtain and then verify the customer's contact details and add them to an SQL database.

To implement this, we need to create a server using fastapi and uvicorn, to manage the requests sent by the chatbot asynchronously. The server deployed by uvicorn is located at http://localhost:8000, but Dialogflow only accepts secure URLs. It was therefore necessary to use a tunnel that generates an https URL.

Then, to extract the relevant information from the Dialogflow responses, all I had to do is study their JSON syntax and filter out what's useful and what's not. Here is what a an extract of fulfillment request look like, with added comments:

```
"responseId": "98eb2642-eea6-4f74-9495-5a2c60118907-47793ac9",
 "queryResult": {
  "queryText": "my phone number is 0123456789 and my email address is
contact@example.com",
  "parameters": {
    "phone-number": "0123456789"
   "email": "contact@example.com" jj #Extraction of important features from the chat
  },
     "text": {
      "text": [
       "Are the following information correct: \nEmail address: contact@example.com\nPhone
number: 0123456789"
      ]
     }
   }
   'outputContexts": [
     "name": "projects/chatbot-health-wmdg/agent/sessions/214c489a-e9f7-bea0-5a90-
e12e7c46e8c8/contexts/check-coordinates", #Extraction of the context, useful to handle
multiple sessions
     "lifespanCount": 5,
     "parameters": {
      "email": "contact@example.com",
      "phone-number": "0123456789", "email.original": "contact@example.com",
      "phone-number.original": "0123456789"
    } ...
  "intent": {
   "name": "projects/chatbot-health-wmdg/agent/intents/d79dc514-8034-4fe9-a7f5-
e554b53319bc",
   "displayName": "check.coordinates" #Extraction of the intent name
  "intentDetectionConfidence": 1,
  "diagnosticInfo": {
   "webhook latency ms": 143
  "languageCode": "en"
 } ....
```

An important step of the development of the backend is handling the multisession: when several users connect to the chatbot simultaneously, each session needs to maintain its own context and data, ensuring personalized and accurate responses without interference from other users. To do so, I created a dictionnary and associate a key for each session ID.

This isolation guarantees a smooth and secure experience, even with numerous interactions running in parallel.

Finally, the last step of the backend was to create a SQL database and establish a connexion with the python script.

The database is really simple. It contains the three following tables:

| # | product_id | product_name |
|---|------------|--------------|
| 1 | 1 | MedIAg |
| 2 | 2 | SyntheMedix |
| 3 | 3 | ForecastMed |
| * | NULL | NULL |

products:

A unique ID (primary key) is associated to each product

| # | customer_id | phone_number | email_address |
|---|-------------|--------------|----------------------|
| 1 | 1 | 0123456789 | contact1@example.com |
| 2 | 2 | 1234567890 | contact2@example.com |
| 3 | 3 | 2345678901 | contact3@example.com |
| 4 | 4 | | contact1@test.com |
| 5 | 5 | 1234567890 | contact2@test.com |

customers:

A unique ID (primary key) is associated to each customer, with their coordinates

| # | customer_id | interests | date |
|---|-------------|-----------|---------------------|
| 1 | 4 | 1 | 05-11-2024 10:42:17 |
| 2 | 5 | 3 | 05-11-2024 11:06:02 |
| 3 | 6 | \ | 05-11-2024 11:12:34 |
| 4 | 7 | \ | 05-11-2024 11:16:35 |

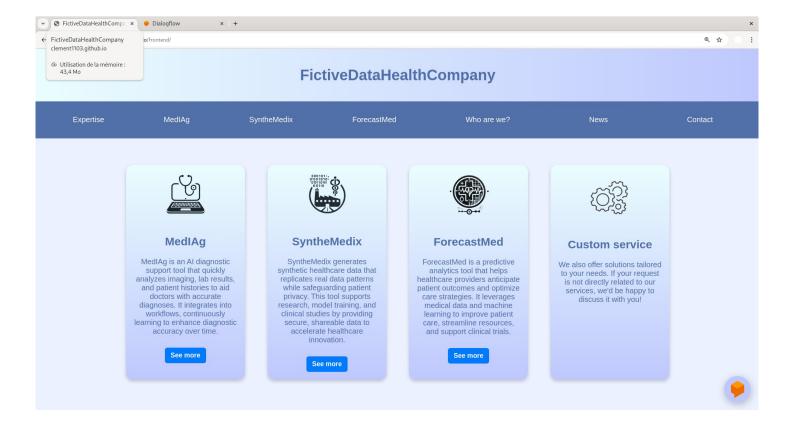
interests:

For each customer (called by their ID as a foreign key from the previous table), we enter the interests and the date of the chat. The column interest contains a str with the ID of each product of interest, or '\' if none has been mentioned

Thanks to the library mysql.connector, the database is automatically completed at the end of a chat with a user.

5. Frontend development

Finally, I wanted to implement a lightweight frontend to support my chatbot. I mainly used ChatGPT to generate index.html and style.css files, resulting in the following interface:



There is a brief presentation of each product, and a navigation banner where the user can click in order to be redirected to other pages (the redirections are not functional).

At the bottom right, there is an icon allowing to open and close the chatbot window.

6. Deployment and tests

The final step of this project was hosting the frontend, the backend scripts and the SQL database online.

- To host the frontend, I used Github pages which allows to host static websites directly from a GitHub repository, making HTML, CSS and JavaScript files accessible online without server configuration.
- To host the backend, I chose Heroku, a cloud platform that makes it easy to deploy, manage and scale web applications by running their backend and associated services.

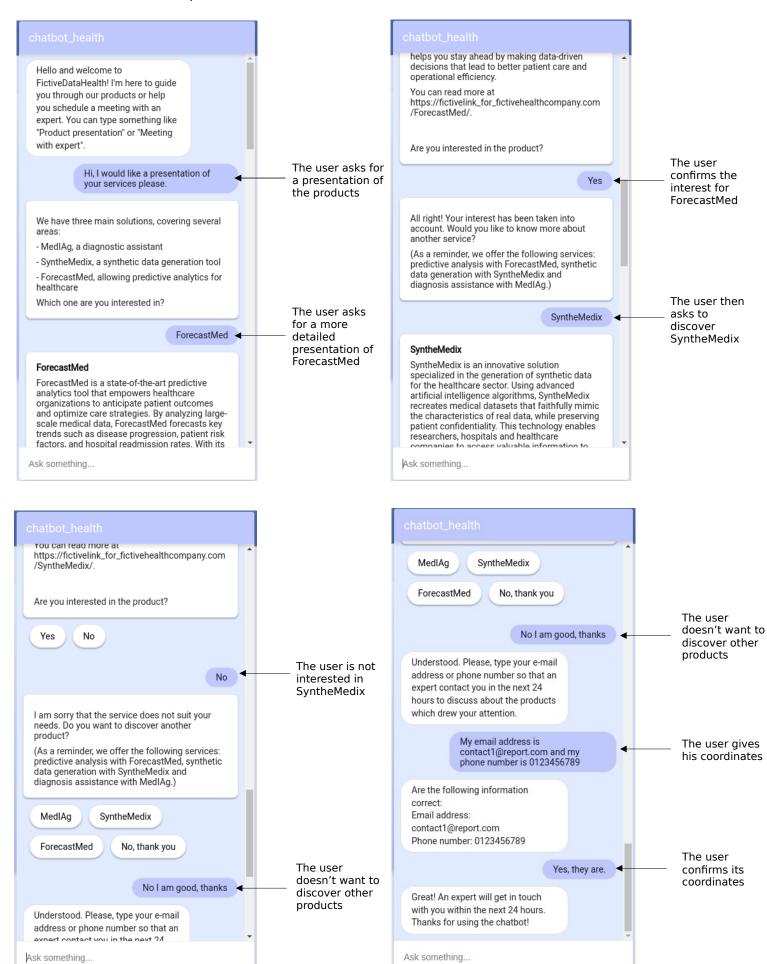
Note: The backend and the frontend codes are present in the repository /data_science_projects/DL_projects/chatbot_health but they are deployed on another github repository.

 For the SQL database, I chose the free plan of AWS RDS which hosts and manages SQL databases in the cloud, automating maintenance tasks such as backups and security. For safety reasons, the database identifiers written in the shared python scripts are not the correct ones.

You can access to the chatbot following this link: https://clement1103.github.io/frontend/

Note: The program works perfectly locally, accessing the database stored on AWS, but I've noticed that the backend sometimes malfunctions when deployed on heroku, due to poor management of the variable corresponding to the different sessions. In fact, the "sessions" variable is stored in volatile memory and is sometimes reset during the chat. I intended to use Redis to improve the storage of variables, but there is no free plan available.

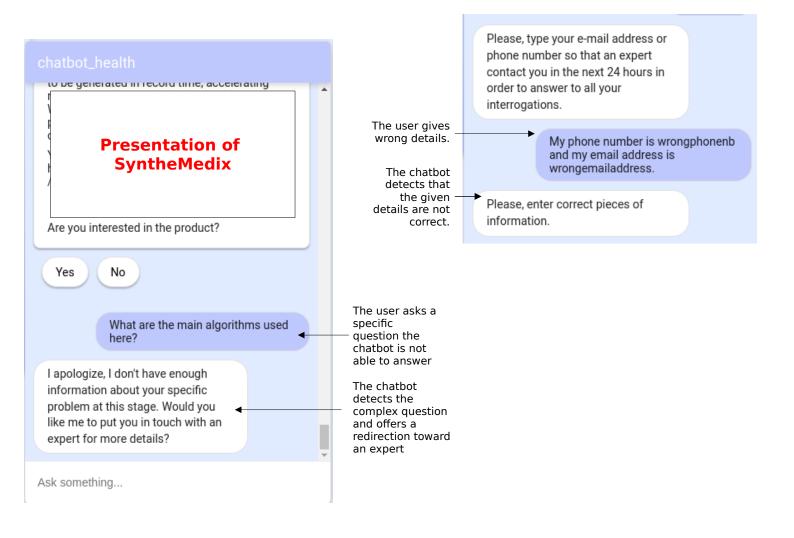
Here is an example of conversation flow:



Following this chat, here are the rows added to the database:

| # | customer_id | phone_nu | mber | email_address | customers table |
|---|-------------|-----------|-----------|---------------------|---------------------------------|
| 1 | 21 | 012345678 | 19 | contact1@report.com | |
| # | customer_id | interests | date | | interests table |
| 1 | 21 | 3 | 05-11-202 | 24 21:19:45 | (3 is the index of ForecastMed) |

Here are some other extracts of other flows, illustrating that the chatbot is polyvalent.



7. Conclusion

In the end, this project was very interesting, as it enabled me to discover the entire tool design process, from specification to production. Indeed, it allowed me to led a serious thinking about the scoping of requirements, and to develop both the backend and the frontend of the project, and manage a light SQL database. Each of these steps allowed me to confront several challenges in the final deployment of an online product.