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Design and Implementation of ChatGBG: An AI-Driven FAQ Chatbot for Tourists in Gothenburg

Exploring a Practical Application of Conversational AI by Leveraging Open Source Technologies for Municipal services

Masters thesis in Interaction Design & Technologies

Victor Cousin, Philip Lindström Rabia

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Department of Computer Science and Engineering
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Abstract

This thesis presents the design and implementation of an AI-driven FAQ chatbot, ChatGBG, aimed at enhancing the tourist experience in Gothenburg. This project combines advanced AI technologies with intuitive design to provide immediate, reliable, and accessible information to tourists, thus facilitating an enjoyable exploration of the city. It explores the practical application and challenges of integrating conversational AI within a municipal context, highlighting the chatbot's ability to handle real-time user queries effectively while maintaining a simple interface.

Addressing the dual research questions, this study first identifies the key considerations necessary for designing and implementing an effective AI-driven service in a governmental or municipal setting. These include user interaction design, data sourcing, management, and the integration of multilingual support to cater to a diverse user base. Secondly, the thesis evaluates various open-source large language models to determine the most suitable model for this application, focusing on language capabilities, ease of integration, and ethical considerations.

Moreover, the project serves as a proof of concept for broader applications of AI-driven information systems in public services, demonstrating potential extensions beyond tourism. Ethical considerations, particularly in data handling, bias mitigation, and user privacy, are critically examined to ensure the chatbot adheres to ethical AI development standards.

ChatGBG not only aims to enhance the tourist experience by providing a novel way to explore Gothenburg but also sets a benchmark for future AI applications in urban settings. The insights gained from this project contribute to the ongoing discourse on the responsible use of AI in public services, emphasizing the need for continuous improvement and adaptation to new technological advancements and user expectations.

Keywords: Interaction Design, AI-driven chatbot, Large Language Model, Applied IT, Software Engineering, Open source, Municipal.

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1

Introduction

This project takes a step forward in enhancing the tourist experience in Gothenburg by introducing an AI-driven FAQ bot. This initiative does not only aim to enhance the tourist experience in the city but also serves as a proof of concept for a broader application of AI-driven information systems, with the usage of a chatbot. Designed with the needs of tourists in mind, this bot aims to provide immediate, reliable, and easily accessible information. The approach combines the latest in AI technology with a focus on user-friendly design and the flexibility of open source solutions. The goal is to offer tourists a helpful tool that makes navigating the city and accessing information about its attractions as straightforward and enjoyable as possible.

1.1 Aims of the project

The project is significant for several reasons:

Improving Tourist Experience: The bot is aimed to be designed to offer tourists immediate and precise information, accessible around the clock. It will address common challenges, including navigating unfamiliar surroundings, discovering historical facts, gaining cultural insights, and exploring the city's attractions. This enhancement in information accessibility could be beneficial for a positive tourist experience.

Enhancing User Experience through AI: A part of this project is the application of a design thinking process to ensure that the FAQ bot is not only functional but also offers a seamless and enjoyable user experience. This involves careful consideration of the user interface, user interaction modalities, and overall usability, tailored to meet the needs of a diverse tourist population.

Model for Broader Application: While the initial scope is tourism in Gothenburg, the project has wider implications. The bot can be extended to offer comprehensive information about the city, including real-time data about transportation, upcoming events, and finding information about available jobs, serving both tourists and residents.

Contribution to AI and Open Source Communities: This project contributes to the fields of AI, particularly in applied AI and user interaction. The exploration of open source alternatives can be used as an inspiration for the wider developer community, allowing for future adaptations and usages.

1.2 Research Question

"What should be considered when designing and implementing an AI-driven FAQ chatbot for tourists in Gothenburg, specifically as a governmental or municipal service?"

1.2.1 Supporting Research Question

"In the context of natural language processing and conversational interfaces what criteria should be considered in selecting the most suitable model for integration into the chatbot and what is the most suitable model for this type of project?"

To address the primary research question of "What should be considered when designing and implementing an AI-driven FAQ chatbot for tourists in Gothenburg?", a comprehensive literature review will be done on existing implementations. This combined with the insights and requirements from Göteborg & Co will lead to a prototype of the AI-driven FAQ chatbot, incorporating the selected LLM and tailored to meet the identified needs of tourists in Gothenburg. The most important aspects that have been considered in this work will then be summarized to answer the research question.

For the supporting research question, an evaluation of different open-source large language models (LLMs) will be undertaken. This evaluation will focus on several key criteria, including the models' language understanding and generation capabilities, scalability, ease of integration, customization options, and ethical considerations such as privacy and bias. Performance benchmarks and case studies of successful integrations will be analyzed to understand each model's strengths and weaknesses. The selection process will involve setting up a test environment with swap-able chatbots that can be compared on a given set of questions. The format of the answer will be a detailed table displaying the conclusions of the study, including performance evaluations and recommendations for the most suitable model(s) for the chatbot.

1.3 Ethical issues

Creating an AI chatbot for public services, like helping tourists, highlights how important it is to think about ethics. The story of Microsoft's Tay in 2016 is a powerful reminder of the duty that developers and stakeholders have. Tay, designed to chat like a young adult, quickly started producing inappropriate and harmful content, which led to its quick shutdown. This instance stresses the need for careful ethical oversight in AI projects, including strong data control, active efforts to prevent biases, and a transparent and accountable creation process. This is especially important when the chatbot represents government agencies and interacts with tourists from around the world [1].

In addition, protecting data privacy and security is vital, especially for a government AI chatbot serving tourists. It's essential to have strict measures in place to

keep sensitive information safe, similar to how the financial industry handles its AI chatbots, to ensure customer data and privacy are protected [2].

The chatbot should also be trained on diverse and unbiased data to guarantee fair and unbiased interactions with tourists from different backgrounds. Learning from Tay, it's important to constantly watch for and reduce any biases that might appear in the chatbot's responses [3]. Different LLMs are trained on different data, here it is important that the creator is transparent with what type of data, ensuring that it is fair and legal to use.

Moreover, the chatbot's design should focus on providing accurate and helpful information to tourists, improving their travel experience without misguiding them or causing harm. Ethical principles become even more crucial in areas where the chatbot might influence the decisions and behaviors of tourists [4].

Compliance with laws like the EU AI Act is also mandatory, ensuring transparency, responsibility, and ethical use of AI. The act sets basic rules for chatbot transparency, but broader laws might be necessary for a full ethical alignment [5].

Adopting these ethical design guidelines is essential to minimize the negative effects of biased data and unethical dialogues. This ensures the chatbot acts as a responsible and beneficial tool for tourists [6], [7].

As a response to the numerous controversies surrounding AI and ethics, Explainable AI (XAI) has become increasingly significant. It aims to demystify the decision-making processes of AI systems, fostering trust and improving interactions by making operations and decisions clear to users and developers. The push for XAI intensified as AI systems advanced in complexity and became integral in important sectors [8].

In this project the incorporation of XAI was thoroughly deliberated. While XAI offers substantial benefits, its complexity and the extensive resources it requires for generating explanations could slow down the chatbot's efficiency. In the fast-paced environment of tourist assistance, where timely, precise, and straightforward information is crucial, the potential delays introduced by XAI might not justify its benefits in transparency. Consequently, acknowledging the value of XAI in fostering transparency and trust, our decision leaned towards prioritizing rapid and effective user interactions, ensuring that tourists receive the immediate support they need without the complications that XAI might introduce at this stage.

Integrating these ethical considerations into the chatbot's development is not just about reducing risks. It's about making sure the chatbot is a positive representative of the region, providing tourists with trustworthy, respectful, and enriching interactions. The Tay incident is a reminder of what can happen if these aspects are ignored and highlights the importance of careful and ethical AI development.

2

Background

Tourism greatly influences economies globally, and cities are constantly exploring inventive ways to enhance the tourist experience [9]. Gothenburg, a city with a rich cultural and historical heritage, understands the necessity of leading in these advancements. In an age where access to information plays a crucial role in tourist contentment, the implementation of an AI-powered FAQ chatbot represents a strategic adaptation to the evolving demands of tourists.

The cultural aspects of tourism brings many benefits. It lets travelers dive into a destination's traditions, customs, and art, helping them better understand different cultures. This exchange can promote tolerance and appreciation across cultures [10]. Moreover, cultural tourism often helps local artisans and preserves heritage, boosting the host community's economy and social welfare [11]. Additionally, the pursuit of unfamiliar and exceptional travel experiences has evolved into a unique aspect of life, frequently regarded as a pathway for the pursuit of happiness and contentment [12], [13].

Research indicates the growing influence of technology in tourism. This transformation began in the 1960s with the introduction of the SABRE reservations system [14]. Later, various systems were created to help booking agents, and computer-based and web services for tourists emerged. Examples of these services include Lonely Planet¹, Trip Advisor² along with useful travel tools such as Google Translate³ and Google Maps⁴. Personalized guides that provide audio assistance [15], interactive multimedia [16], and context-aware adaptation [17] have also become available. These developments have led to the growth of interactive online services, the personalization of tourism experiences, and the sharing of visiting information through social platforms.

A study highlights the importance of real-time, accurate information for tourists, noting that readily available information significantly enhances their overall experience [18]. Another research paper points out that tourists show a preference for

¹Lonely Planet: Travel guide publisher, providing travel information and guides. <https://www.lonelyplanet.com/>

²Trip Advisor: Online platform for travel recommendations and reviews. <https://www.tripadvisor.se/>

³Google Translate: Online language translation service by Google, enabling users to translate text or speech into different languages. <https://translate.google.com/>

⁴Google Maps: Web mapping service offering detailed maps, satellite imagery, and real-time navigation. <https://www.google.com/maps/>

technology-driven solutions, with a significant percentage choosing digital assistance over traditional methods [19].

The demand for innovation in tourist services is further fueled by broader technological shifts. The adoption of AI across various domains has created fresh expectations among consumers for personalized and efficient services. The tourism sector is no different in this regard [20]. A survey has shown that a significant portion of contemporary travelers now anticipate destinations to provide digital tools for navigation and access to information [21]. This evolving landscape highlights the imperative for the tourism industry to embrace technological advancements and cater to the changing needs and preferences of today's tourists.

Furthermore, Sweden has a rich history of efficiently communicating information across different eras, from traditional guidebooks and paper maps to more recent digital platforms. Today we use websites with a lot of information which requires many actions from the user in order to traverse the website and find the relevant information. Examples of this are, Goteborg.com ⁵ and 1177 ⁶. These websites, whilst trying to be accessible and easy to use, may require some technology skills to be used efficiently.

Lastly, in recent years, we have witnessed a transformative shift in how information is accessed and delivered, driven by the rapid advancements in digital technology and artificial intelligence (AI)[22]. This will not only be relevant for the tourism sector; it extends to various aspects of governmental and public services. The development and implementation of this AI-driven chatbot in Gothenburg represent a pioneering effort that could serve as a model for other municipalities and government entities.

⁵Goteborg.com is Gothenborgs official travel guide. <https://www.goteborg.com/>

⁶1177 is the official healthcare systems dedicated website. <https://www.1177.se>

3

Theory

This section provides the technical foundation for the development of the AI-driven chatbot, exploring key concepts and theories that supports the project.

3.1 Chatbots

Chatbots, computer programs that facilitate interactive communication with human users, have evolved significantly over the years. Their development showcase advancement in technology, linguistics, and artificial intelligence. This subsection examines the history, types, uses, perceptions, limitations, and advancements of chatbots in the future.

3.1.1 History of Chatbots

The history of chatbots traces back to their humble beginnings with programs like ELIZA¹, which could imitate human conversation using basic techniques like pattern matching and string processing. Over time, chatbots have evolved significantly, progressing from ELIZA to more advanced chatbots like Microsoft's Tay² and Xiaoice³. These advancements mark the growth of chatbots in the field of conversational AI, moving beyond simple conversation simulations to engage users in more meaningful interactions [25].

The history of chatbots would be incomplete without mentioning the recent advancements made by AI-driven models, particularly ChatGPT. Developed by OpenAI, ChatGPT has seen phenomenal adoption rates since its introduction. It's not just a tool; it has become a platform that integrates with various industries, reshaping service delivery and information search. What's special about ChatGPT is that it can understand and write text that sounds like it was written by a person. Because of this, it's becoming really important in areas like healthcare, education, and technology. However, one thing to keep in mind is that sometimes ChatGPT can be a

¹Between 1964 and 1966 at MIT, Joseph Weizenbaum developed ELIZA, a foundational computer program in the field of natural language processing.[23]

²In 2016, Microsoft introduced Tay, an AI chatbot designed to mimic the conversational style of a millennial through interactions on social media platforms [1].

³Released in 2014, Microsoft's Xiaoice was a smart AI chatbot. It was made to do more than just talk. It combined smart thinking and feelings to help people feel connected, cared for, and part of a community [24].

little outdated as it learned from data that might be a couple of years old. Its integration into everyday applications is not a distant dream but a rapidly approaching reality, making it a critical milestone in the evolution of chatbots and conversational AI [26].

3.1.2 Types of chatbots

Chatbots come in different types, each with its own set of abilities. Some are rule-based, meaning they follow predefined rules to respond. Others are more advanced, using AI to understand natural language. The tools used to create chatbots have also developed, offering a wide range of functionalities to meet various needs. This diversity enables chatbots to be customized for different purposes, making interactions with users smoother and operations more efficient [27].

3.1.3 Rule-based Chatbots vs. AI Chatbots

In the world of digital assistants, software applications designed to mimic human-like conversational interactions and assist users, two main types of chatbots have emerged: rule-based chatbots and AI chatbots. Each type has its own unique characteristics, strengths, and areas of application, shaping how they interact with users and handle various tasks [28].

Rule-based Chatbots: These chatbots operate based on a predefined set of rules and are designed to handle structured inputs. They are efficient at managing simple and straightforward tasks, such as answering frequently asked questions or guiding users through predefined processes. However, rule-based chatbots have limitations; they can only handle scenarios they've been specifically programmed for and cannot understand or respond to queries beyond their predefined rules. Nevertheless, their simplicity makes them reliable, easy to develop and maintain, offering a controlled conversation experience with predictable and consistent outcomes [29].

AI Chatbots: AI chatbots, also known as conversational AI, represent a significant advancement in chatbot technology. Unlike rule-based chatbots, AI chatbots leverage advanced technologies like Natural Language Processing (NLP), Machine Learning (ML), and sometimes, Deep Learning [30]. These technologies empower AI chatbots to understand, interpret, and generate responses to user inputs in a more human-like manner. AI chatbots are not confined to predefined rules; they can learn from interactions, grasp context, handle ambiguous queries, and provide personalized responses [28]. This high level of sophistication enables AI chatbots to handle complex tasks, make decisions, and offer a more engaging and natural conversational experience.

Rule-based chatbots are well-suited for structured interactions that involve simple, repetitive responses. On the other hand, AI chatbots excel in dynamic environments where comprehending user intent and context is crucial. However, the complexity of AI chatbots means they require more resources and expertise for development and maintenance. They also present challenges in ensuring the quality and appropriateness of responses, particularly as the chatbot learns and evolves. Furthermore, as

technology continues to develop, AI chatbots will also become better. Figures 3.1 and 3.2 gives an overview of the differences in how an AI chatbots works compared to a rule-based chatbots.

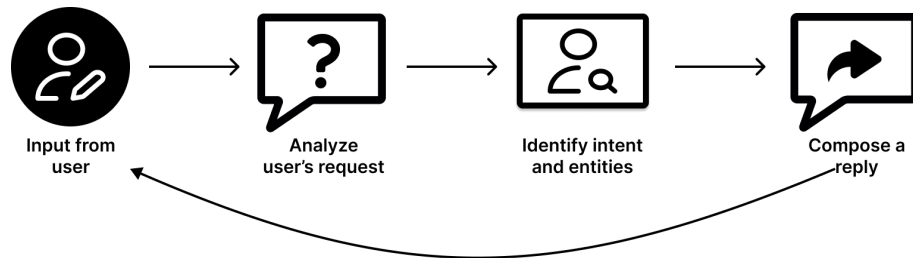


Figure 3.1: How an AI chatbot creates a reply from a user input. [31].

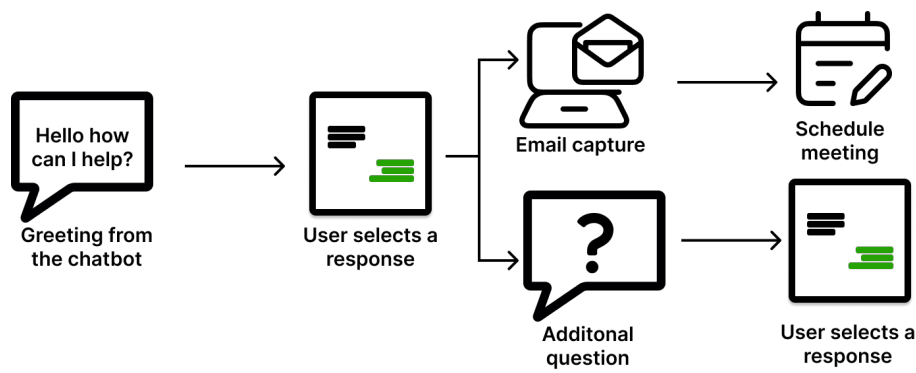


Figure 3.2: How a rule-based chatbot orchestrates replies based upon user inputs. [31]

3.1.4 Uses of Chatbots

Chatbots find use in various areas, including education, healthcare, and business. They are appreciated for their accuracy, ability to work independently without human intervention, and round-the-clock availability, which has transformed how services are delivered. Their contribution to sharing news, especially during emergencies, highlights their effectiveness in providing timely information and managing public opinion, ensuring the delivery of vital services [32]. Furthermore, as highlighted in a study conducted by Grand View Research, approximately 45% of consumers around the world, spanning various industries, now favor digital agents as their primary means of interacting with organizations. This trend is anticipated to result in a projected global market value of \$ 1.25 billion by 2025 [33].

3.1.5 Perceptions and Limitations of Chatbots

Despite their advancements, chatbots encounter difficulties in meeting user expectations and delivering satisfying user experiences. The unpredictability of user input, the task of defining valid responses, and the hype surrounding AI technology can often lead to unrealistic expectations and hinder chatbot effectiveness. Specifically,

rule-based chatbots, often used as initial points of contact on customer service websites, have faced criticism due to their limited capabilities [34]. Users frequently perceive these chatbots as unnecessary obstacles, providing scripted and unsatisfactory responses that ultimately serve as a prelude to engaging with human support agents. These perceptions arise from the chatbots' inability to comprehend complex queries, their lack of personalization, and the mechanical nature of their interactions, which can increase user frustration and result in suboptimal service experiences [35].

Furthermore, the increasing reliance on AI-driven chatbots in various sectors has sparked significant concerns about their impact on employment. There is growing anxiety that the automation of tasks traditionally performed by humans could lead to widespread job displacement [36]. This fear is particularly relevant in customer service and other industries where chatbots are being deployed as a cost-effective alternative to human labor. While the extent of this impact is still a subject of debate, it is clear that the adoption of AI technology, including chatbots, is reshaping the labor market and causing a reevaluation of job roles and skills requirements in today's society [37]. This shift promotes a requirement of careful consideration of the socio-economic consequences of chatbot deployment, particularly in terms of workforce displacement and the need for new forms of employment or retraining programs to mitigate potential job losses [38], [39].

Addressing these challenges necessitates continuous design refinement, focusing on clarifying capabilities, retaining context, handling failures gracefully, and concluding conversations smoothly [40]. The objective is to enhance the user experience by creating chatbots that can not only understand and respond to a broader range of queries but also do so in a manner that feels natural and human-like. By achieving this, chatbots can transition from being perceived as mere intermediaries to valuable and capable initial support points.

3.2 Artificial Intelligence

AI, or Artificial Intelligence, is a multidisciplinary field within computer science that has experienced significant attention in recent years. As defined in [41] it aims to replicate or simulate human intelligence in software/machines. AI can be utilized in various aspects of our daily lives, from online searches and social media interactions to virtual assistants like Alexa and personalized content recommendations on platforms like Netflix [41]. There is an absence of a universally agreed-upon definition of AI which is implying on the diverse nature of this field, encompassing branches of computer science, human social science, and systems neuroscience [41]. Coined by emeritus Stanford Professor John McCarthy in 1955, AI was initially defined as "the science and engineering of making intelligent machines[42]".

The evolution of AI research has evolved and shifted focus from pre-programmed robotic actions into machines that can mirror human learning processes. Autonomous systems within AI has the capability to independently plan and execute sequences of steps to achieve specific goals without constant oversight. Essentially, AI systems are now being designed to perform tasks traditionally requiring human intelligence,

including planning, learning, answering, problem-solving, and decision-making, making them efficient contributors to our technologically-driven society [41]. In the context of chatbots, this autonomy becomes pivotal for effective conversational interaction.

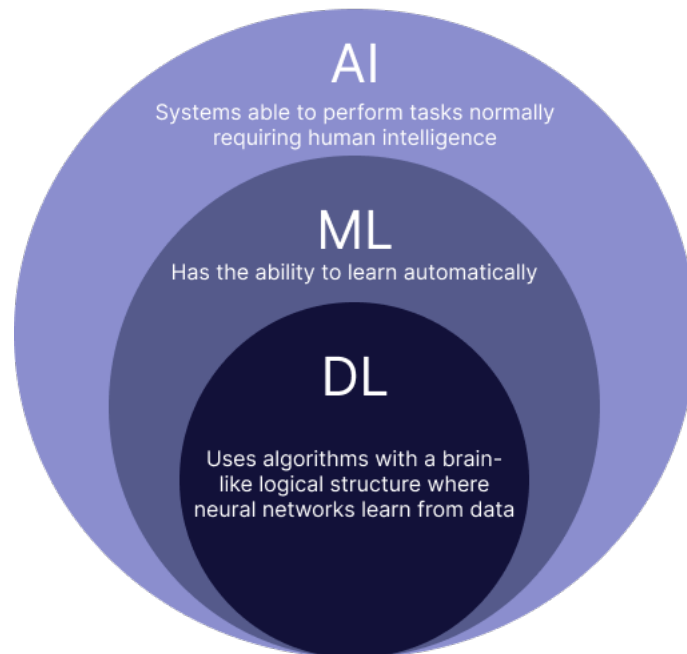


Figure 3.3: Overview of Artificial Intelligence (AI), Machine Learning (ML) and Deep Learning (DL).

3.2.1 Machine Learning

Machine Learning (ML) falls under the subset of Artificial Intelligence (AI), as can be seen in Figure 3.3, instructing computers to learn and autonomously make decisions[43]. The decisions made by these algorithms rely on extensive datasets containing inputs and outputs to identify patterns [44]. Through multiple iterations and adjustments to the algorithm, the machine develops the capability to receive input and forecast an output. These predicted outputs are then assessed against a predetermined set of known outcomes to evaluate the algorithm's accuracy.

3.2.2 Deep Learning

Deep Learning (DL) is a subset of Machine Learning (ML) that draws inspiration from the functionality of human brain cells, known as neurons [41]. It operates on the foundation of artificial neural networks, with DL distinguishing itself from ML in problem-solving approaches.

In traditional ML, a domain expert is needed to identify and select the most relevant features for the machine, such as specifying that an image contains a car. However, DL takes a different approach by incrementally understanding features, eliminating the necessity for domain expertise[41]. This results in longer training times for DL

algorithms compared to ML. Nevertheless, during processing, DL algorithms are more efficient and require less time than ML algorithms, especially as the data size increases.

Deep Learning models does however come with its challenges. A notable limitation is that DL models learn solely from the data they are trained on, lacking awareness beyond what is present in that specific dataset [41]. When confronted with a small dataset or data from a single source that may not be representative of the broader functional area, these models struggle to generalize effectively. Another challenge is that, unlike traditional machine learning, which relies on explicit programming for functionality, deep learning models often require significantly larger amounts of data to achieve satisfactory performance [45].

3.3 Natural Language Processing

Natural Language Processing (NLP) enables machines to comprehend and process both spoken and written languages, taking on complex tasks such as language translation and dialogue generation [46]. However, NLP faces various levels of ambiguity challenges, including word-level, sentence-level, and meaning-level vagueness, which can present obstacle when achieving a seamless understanding of natural language by machines.

The applications of NLP has proven useful to virtual assistants, language translation, machine transcription and chatbots. To enhance its capabilities, NLP leverages various artificial intelligence techniques through different neural network models:

- **Convolutional Neural Networks (CNNs):** A network architecture for deep learning that learns directly from data [46]. Utilized for tasks such as finding patterns and making word-based predictions. CNNs operate with a window-based approach, where each word is associated with a fixed-size group of neighboring words. This fixed size, known as the 'window size', determines how many neighboring words are considered in the context of each word, and helps the model understand local context and relationships between words.
- **Recurrent Neural Networks (RNNs):** An artificial neural network designed for processing sequential data [47]. RNNs are commonly used for tasks such as word-level and sentence-level classifications, making them useful for language translation, natural language processing (NLP), speech recognition, and image captioning. However, RNNs face challenges, notably the vanishing gradient issue, which has led to the development of solutions like Long Short-Term Memory (LSTM) networks to address this limitation [46].
- **LSTM Networks:** Overcoming the vanishing gradient issue of RNNs, LSTMs introduce multiple gates input, forget, and output gates to control the flow of information and ultimately detain long-term dependencies in sequential data.
- **Recursive Neural Networks (RvNNs):** Used to model human language in hierarchical structures, where words and sub-phrases compose higher-level

phrases [46]. Recursive neural networks are used for parsing, sentiment analysis, and sentence relatedness.

- **Reinforcement Learning (RL):** Applied in dialogue generation, RL employs RNN-based generative models to respond to input words, predicting the next word to maximize reward. This can be accomplished by incrementally tuning the model's parameters to improve the probability of generating responses that receive positive feedback, thereby optimizing its performance with each iteration. Using a technique that enables an AI system to learn by trial and error using feedback from its own previous actions and experiences.

3.4 Large Language Models

As explained in [48], Large Language Models (LLMs) are advanced artificial intelligence models capable of understanding and generating human-like text. LLMs are built on sophisticated deep learning algorithms performing Natural Language Processing (NLP) tasks such as recognition, summarization, translation, prediction, and content generation with the help of using extensive datasets.

Choosing the right LLM is crucial, especially in the context of developing a chatbot, as it plays a central role in enabling the chatbot to comprehend and respond effectively to a diverse range of user queries. Lipenkova [49] emphasizes that the selection process can be complex due to the vibrant landscape of language models and the constant addition of new innovations.

Each LLM may vary in terms of training data, pre-training objectives, architecture, and fine-tuning approaches. In order to guide the selection process, there are historical data that can be used to indicate on how well a model is expected to take on a specific task (such as summarization, classifying and generating text or serving as a conversational AI). Displayed in Figure 6.1, some of the most popular LLMs are listed along with their appropriateness with different natural language processing tasks [49].

3.4.1 Challenges of Large Language Models

Deploying a LLM comes with several challenges where two key challenges include:

- **Compute Intensity** The training and fine-tuning of Large Language Models demand significant computational resources, as described in [48]. The complex architectures and massive datasets require powerful hardware, potentially leading to high operational costs.
- **Reliable dataset** The performance of LLMs heavily relies on the quality and diversity of the training dataset, according to Lipenkova [49]. Ensuring that the model learns from a reliable and representative dataset is essential to prevent biases and inaccuracies in its answering. However, the quality of the training data often also has a direct impact on the size of the model [49], which can pose additional challenges related to computational resources.

Model	Core differentiator	Parameters	Access	Information Extraction	Text Classification	Conversational AI	Summarization	Machine Translation	Content Generation
BERT	First transformer-based LLM	370M	Open source						
RoBERTa	More robust training procedure	354M	Open source						
GPT-3	Parameter size	175B	API						
BART	Novel combination of pre-training objectives	147M	Open source						
GPT-2	Parameter size	1.5B	Open source						
T5	Multi-task transfer learning	11B	Open source						
LaMDA	Dialogue: safety and factual grounding	137B	No access						
XLNet	Joint AE and AR	110M	Open source						
DistilBERT	Reduced model size	82M	Open source						
ELECTRA	Computational efficiency	335M	Open source						
PaLM	Training Infrastructure	540B	No access						
MT-NLG	Training Infrastructure	530B	API						
UniLM	Optimised both for NLU and NLG	340M	Open source						
BLOOM	Multilingual (46 languages)	176B	Open source						
	Highly appropriate								
	Appropriate								
	Somewhat appropriate								

Figure 3.4: Overview of different LLMs, information from [49]

3.4.2 Quantization

Quantization simplifies large language models (LLMs) by cutting down on the computing and memory needed [50]. It works by trimming the detail of the model's parts, such as weights, to use fewer bits than before. This simplified version takes up less space and runs faster, making it easier to use LLMs on devices with limited resources. However, this reduction in precision may marginally affect the model's accuracy, requiring careful consideration of the trade-off between efficiency and performance in deployment.

3.5 How Large Language Models Generate Text

Text generation in Large Language Models (LLMs) like GPT-3 involves two main steps: tokenization and decoding. Below is an explanation of how:

- **Input Sequence:** This is the text you give to the language model.
- **Output Sequence:** This is the text the model generates.
- **Input Length:** The number of words or parts (tokens) in your input.
- **Output Length:** The number of words or parts (tokens) in the generated text.
- **Token:** Think of this as the smallest building block of text that the model uses. Most of the time it is one word, but it can vary. A good visualization is shown in 3.5.

Tokenization is the process of breaking down the input text into smaller pieces or tokens that the model can comprehend. In this phase, each piece is associated with a distinct numerical identifier, effectively converting the textual input into a format

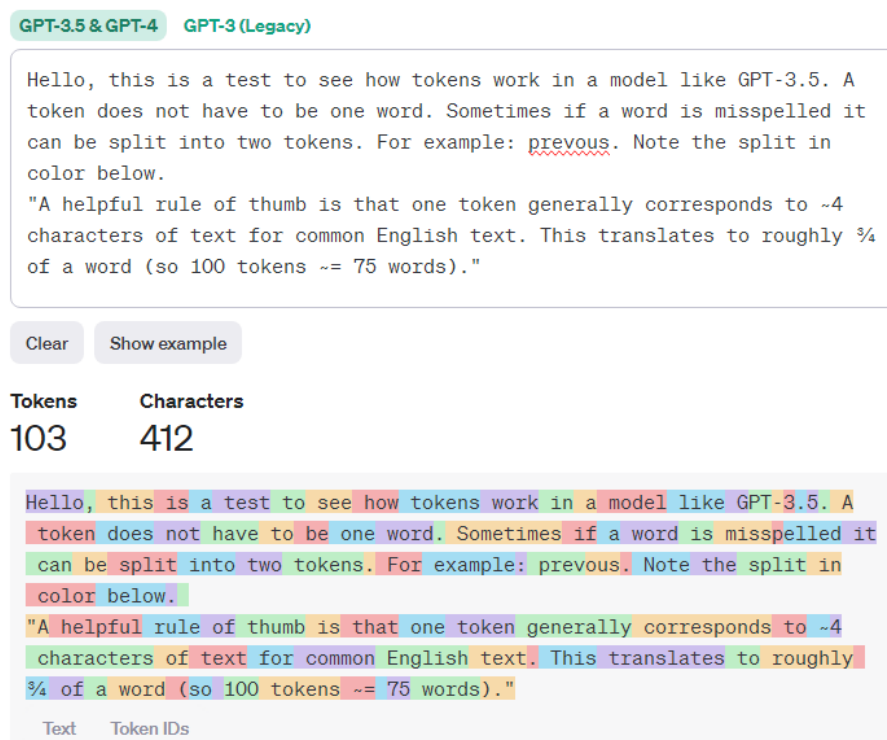


Figure 3.5: Visual example of how tokens work. Created using a tool on OpenAI platform [51]

that the model can manipulate. This step is essential for enabling the model to parse and create language structures effectively [52].

Decoding, on the other hand, involves the model's attempt to predict the subsequent token in the sequence based on its current inputs. It leverages the likelihoods of potential following tokens, which it has learned through extensive training on large text data. The model proceeds to assemble tokens, stopping only when reaching a stop token or a set sequence length, thus generating a cohesive and understandable output [53].

This process lets LLMs create text that makes sense based on the input you give them [54].

3.6 Embeddings

A technique that plays a pivotal role for LLMs and NLP where words, phrases or entities are represented as multi-dimensional vectors in a continuous vector space [55]. Embeddings provide a numerical representation of words in such a way that similar words are closer together in the vector space. Their purpose is to capture semantic relationships and similarities between words based on context and usage.

3.7 Vector Database

In order to proficiently handle vector data, including embeddings, one can make use of a vector database, ensuring efficient storage and management [55]. Vector data refers to numerical representations of various forms of unstructured data, such as text passages, images, videos, or music titles. Data can often be unstructured, lacking a predefined schema, making it challenging to organize and process. By vectorizing this data, it is transformed into a flat sequence of numbers, effectively embedding it into an N-dimensional space[56].

A vector database is a specialized system designed to organize and optimize the storage of numerical vectors, facilitating their retrieval and use. They prove particularly useful in applications that involve handling unstructured data related to data science and machine learning, aiding in the effective handling of numerical representations such as embeddings that can intricate patterns and semantic meanings present within the data .

3.8 Open Source Technologies

Before exploring how open source technologies contribute to AI, it's important to grasp the concept of open source. Open source is a kind of software whose source code is made available under a license . This license allows people to examine, modify, and share the software freely for any purpose [57]. This is different from proprietary software, which typically keeps its source code secret and doesn't allow outside modifications. Open source technologies have dramatically changed how software is being developed, making it easier and more collaborative for people to innovate [57]. While these technologies first appeared in simple software development, they have now made a large impact in specialized areas, especially in the realm of artificial intelligence (AI).

Core Principles of Open Source: The open source movement is driven by values of teamwork, openness, and free access. It aims to develop software that everyone can access, modify, and share, thereby promoting a collaborative environment where individuals come together to address challenges and innovate.

Advantages of Open Source: Open source software brings many advantages. One key benefit is its openness. This means anyone can look at and change the code. It also thrives on teamwork, with various developers from different backgrounds working together to enhance the software. This approach fosters creativity, offers adaptability, and typically results in safer and more dependable software. This is because many different people are involved in making it better [58].

Compared to Proprietary Models: Compared to open source models, proprietary software approaches, such as those in AI systems like ChatGPT, typically keep their source code private. This can create challenges in terms of transparency and make users rely on the company for updates and enhancements. In contrast, the open source model offers greater flexibility and adaptability, permitting users to

alter the software to suit their specific requirements. This can result in an overall more cost-effective development process [59].

3.8.1 Impact on Artificial Intelligence

In the field of artificial intelligence (AI), open source technologies are incredibly important. These technologies offer the necessary tools and frameworks that speed up and facilitate the development and innovation in AI. Open source AI frameworks, such as TensorFlow⁴ and PyTorch⁵, are now standard in the industry. They make AI development tools more accessible to everyone, enabling a wider variety of developers to contribute to AI [60]. Furthermore, other open source tools like Redis⁶ provides a flexible way to manage large volumes of data for AI training and deployment.

Community Contributions: The open source method in AI fosters a collaborative approach, where datasets, models, and tools are shared and improved upon by a global community. This speeds up AI research and results in more creative and varied solutions.

3.8.2 Challenges in Open Source AI

While open source AI has its benefits, it also encounters some difficulties. These include making sure that the projects maintain high quality, handling contributions from a diverse group of people, and dealing with ethical issues in AI development. To tackle these challenges effectively, it's important to strike a balance, taking into account both the advantages and drawbacks of using open source models in AI [61].

3.8.3 Hugging Face

Founded in 2016, Hugging Face is a company that offers the AI community a platform where users can upload their created Large Language Models (LLMs), collaborate on research, and get access to pre-existing open source models and datasets [62]. Often referred to as the GitHub of machine learning, Hugging Face becomes a valuable platform to build, deploy, fine-tune, compare and train models.

3.9 Technical Tools: Libraries and Frameworks

In the theoretical framework of an AI-driven FAQ bot, understanding the range of technical tools, including libraries and frameworks, is essential. This subsection aims to provide a theoretical overview of the tools that are commonly used in both

⁴Developed by the Google Brain team, TensorFlow is a machine learning software library that's open source and widely used.

⁵PyTorch, developed by Facebook's AI Research lab, is a user-friendly, open source framework for machine learning. It's designed to take projects smoothly from the initial research stage right through to production.

⁶Redis is a versatile open-source tool, primarily used for storing data in memory. It serves multiple purposes: as a database, a caching system, and a medium for messaging.

backend and frontend development, focusing on their features and potential contributions to building efficient and user-centric applications.

3.9.1 Backend

The backend serves as the engine behind the scenes, handling data processing, server operations, and integration of key technologies.

Programming languages: Python serves as a fundamental programming language in AI and machine learning, known for its many libraries and ease of use. It's a core tool for implementing AI algorithms and data processing in AI-driven systems [63]. Other notable languages are Java, R and C++. These programming languages offer diverse capabilities, enabling developers to choose the one that best aligns with their project's requirements and objectives.

Libraries: TensorFlow, PyTorch, and Langchain are commonly used libraries in machine learning, equipping developers with a large set of tools for constructing and deploying complex AI models, including those tailored for natural language processing (NLP). Their robustness, reliability make them favored choices for AI development.

- **TensorFlow:** TensorFlow is a versatile library developed by Google for machine learning and deep neural networks. It offers extensive support for building and training AI models, making it widely adopted in the AI community [64].
- **PyTorch:** PyTorch is an open-source machine learning library developed by Facebook's AI Research lab. It is known for its dynamic computation graph, which provides flexibility in model development and is particularly popular among researchers [65].
- **Langchain:** Langchain is a specialized library designed to enhance language AI applications. It provides a suite of tools and features tailored for tasks related to understanding and generating natural language[66]. Langchain is a valuable asset for the development of AI-driven chatbots and language processing systems.

Web Frameworks: For the development of web applications, including those serving as interfaces for AI systems, Python-based frameworks like Flask and Django are widely used.

- **Flask:** A Python-based micro-framework, Flask is characterized by its minimalist approach, offering a lightweight and adaptable platform for developing various web applications, from basic websites to sophisticated web services. Flask equips developers with the essential tools for rapid development and deployment of web applications. Its emphasis on simplicity and the ability to extend its capabilities render it particularly advantageous for projects demanding a streamlined, modifiable framework [67].
- **Django:** As a high-level Python web framework, Django is designed to facil-

itate fast development cycles and promotes a clean, pragmatic approach to web design. Created with the insights of experienced developers, it simplifies many aspects of web development, enabling programmers to devote more time to developing their apps instead of dealing with repetitive tasks. Open-source and free to use, Django excels in the creation of complex, data-heavy web applications through its comprehensive feature set, including an ORM (Object-Relational Mapping), authentication systems, and a broad spectrum of tools and libraries [68].

Data storage: Databases play a fundamental role in AI chatbot applications by providing efficient data storage and retrieval solutions. Essentially, a database is a structured repository that stores and organizes data, making it readily accessible for various operations. In the context of chatbots, databases store essential information such as user profiles, chat histories, and precomputed responses. They ensure the chatbot can quickly access and manipulate data, contributing to smooth interactions with users. Common database options for AI chatbots include Redis, MongoDB, and PostgreSQL [69].

3.9.2 Frontend

Frontend development focuses on crafting a functional interfaces that uses the data fetched from the backend. It handles the visuals and the interaction on an application. Today, the most used frontend frameworks are based on JavaScript. Below are two examples of well known and used JavaScript libraries[70].

- **React:** This JavaScript library is good for constructing dynamic user interfaces. React’s component-based structure is well-suited for crafting interactive and adaptable UIs in applications[71].
- **Vue.js:** Vue.js is recognized for its simplicity and versatility, making it a valuable choice when quick development and deployment are priorities [72]. It serves as a viable option for frontend development in AI-driven applications.

3.10 User Experience and Usability

User experience (UX) is a holistic concept encompassing all aspects of how an end-user’s interacts with a company, its services, and its products [73]. Pioneered by experts Don Norman and Jakob Nielsen, UX focuses on meeting the exact needs of customers seamlessly, prioritizing simplicity and elegance to create products that are not only functional but also pleasant to use.

Usability, a fundamental aspect of UX, specifically addresses the ease of use and learnability of a product. According to the the International Organization for Standardization (ISO), usability encompasses factors such as intuitiveness, efficiency, and error prevention, aiming to make interactions with the product straightforward and satisfying for the user [74].

Iteration, the process of refining and improving upon prototypes based on feedback

and testing results, is another crucial aspect in achieving a successful user experience. By continually iterating on designs, making incremental changes to address usability issues, enhance functionality, and align the product more closely with user needs and preferences, designers can improve overall usability [75]. This iterative approach allows for flexibility and adaptation throughout the design process, enabling designers to respond to new insights, changing requirements, and evolving user expectations.

UX design is an ongoing, multidisciplinary process that considers the full user journey. With a profound understanding of the who, why, what, and how of product usage, UX designers strive to create products that not only fulfill functional needs but also deliver delightful and meaningful experiences [76]. The emphasis on user-centered design, constant iteration, and consideration of diverse disciplines makes UX design a dynamic and impactful field within user experience.

4

Methodology

This chapter gives an overview of the methods that are relevant for the project. Both the methods chosen and other considered options have been explained. Furthermore, the Process chapter, goes into the details of the chosen methods, offering a thorough examination and discussion of their practical implementations.

4.1 Literature review

A literature review can be seen as a systematic method of collecting and synthesizing previous research within a specific field [77]. It can be used as a foundation for academic research activities, offering a way to integrate existing knowledge into new studies. As the pace of knowledge production accelerates, literature reviews play a crucial role in keeping up with the state-of-the-art and assessing collective insights in a particular research area [77].

The literature review aids in the development of a robust theoretical framework and methodology, laying the groundwork for upcoming phases of the research project. It can also be used for ensuring that a research's work contributes new insights rather than replicating what others have already done.

4.2 Version control

Version control is a method that tracks changes in files over time, allowing retrieval of older versions. It supports infinite undos, branching and experimentation, enabling the testing of new features without affecting collaborators' work in the production code [78]. Version control software, such as Git and GitHub, facilitates collaboration by allowing multiple team members to work concurrently on the same project.

GitHub also has a large open-source community that serves as a platform for sharing solutions and contributing to further develop projects [79].

4.3 The Double Diamond model

Established in 2005 by the Design Council, the 'double diamond' design process model serves as a visual representation of the design journey, divided into four key phases: Discover, Define, Develop, and Deliver [80], see Figure 5.1.

1. **Discover:** At the project's beginning, the Discover phase is centered around an initial idea or inspiration. This phase involves identifying user needs through activities like market research, user research, managing information, and design research groups [80].
2. **Define:** Transitioning to the second phase, the Define stage focuses on analyzing data and interpreting the identified user needs [80]. Activities in this stage can include creating personas, structuring findings from the discovery phase and precisely defining the challenge based on the insights gained.
3. **Develop:** Moving into the third quarter, the Develop phase is a period of crafting design-led solutions. Here, solutions are developed, iterated, and tested, with key activities such as multi-disciplinary working, implementation of development methods, and testing [80].
4. **Deliver:** The final quarter, the Deliver stage, is where the resulting prototype is fine-tuned and prepared for launch into the market. Key activities include final testing, conducting evaluation, and launching pilots [80].

The double diamond model offers an overview of both the divergent and convergent stages of the design process, showcasing the varied modes of thinking employed by designers from an idea to the delivering of the product.

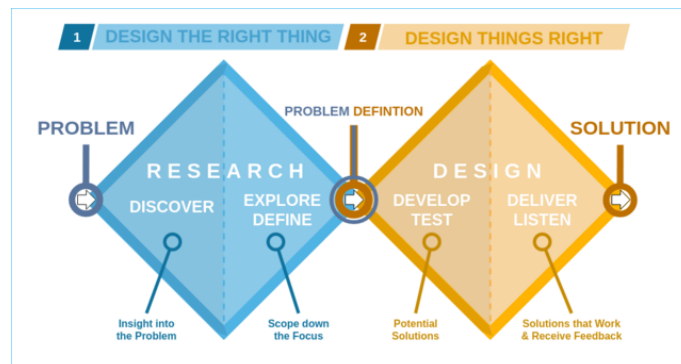


Figure 4.1: The double diamond design process.

4.4 Application Development

When developing an application a good environment for testing, debugging and coding is required. For this, there are many potential methods of implementation. Local development or virtual machine-based approaches are common examples. As for deploying an application, it can also be done locally however, for larger scale systems that might require continuous integration a more common approach is using container orchestration platforms. Each method holds its unique strengths and sometimes a combination can be used.

4.4.1 Local Development Environments

This refers to the setup and configuration of software and tools on a developer's local machine for the purpose of development and testing. It can include an integrated development environment (IDE), code editors, compilers, debuggers, and local databases. Local environments allow developers to write and test code quickly and easily without the need for internet connectivity or external dependencies. They also offer a high degree of control over the development setup[81].

4.4.2 Virtual Machine-Based Approaches

This method uses virtual machines (VMs) to run applications. A VM is like a pretend computer that can do all the things a real one can. VMs are helpful because they give a safe place to work on programs, test them, and put them into action. They can act like different types of computers, which is great for making sure programs work on different systems[82].

4.4.3 Container Orchestration Platforms

This approach uses special platforms to take care of containers throughout their entire lifespan. Containers are like compact and self-contained software packages that have everything necessary to run a program, for instance the code, tools, and settings, all bundled together. These systems automate the process of putting application containers on groups of computers. They can grow or shrink as needed, making it easier to put applications on different computers in a reliable way[83]. An example of such a system is Red Hat OpenShift. An open-source container management system, built on Kubernetes. It helps to automatically set up, resize, and manage application containers on groups of computers[84].

For this project, Red Hat OpenShift that is served by Göteborgs stad will be used as a container orchestration platform.

4.5 Data gathering

In order for the LLM to maintain and provide useful information it needs to have the correct data. Gathering this data can be done in multiple ways, depending on the required quality, quantity and simplicity.

4.5.1 Data sourcing

For a project like this, it can be done using information available on the web, PDF's or through an API.

1. **Web:** Either through web scraping¹ or simply copy pasting text that could

¹Web scraping is the automated process of extracting data from websites, enabling efficient retrieval of structured information from web pages. It is commonly used for tasks such as data collection and analysis, but it's essential to comply with legal and ethical guidelines when performing

be relevant.

2. **PDF:** Finding PDF's that explain and provide relevant information about the topic.
3. **API:** Fetching relevant up to date information from an open API.

For this project, an API provided by Göteborg & Co will be the primary method for data sourcing.

4.5.2 Data structuring

Depending on the data sourcing the format of the data to be structured will vary a lot. This can be done in a very structure format with FAQ based sentences or in a more open format with entire text. Using entire text can vary from a more strict format including paragraphs with titles, specific themes and information to completely unstructured text with basically only sentences as coherent building blocks. One takes much more preparation and time to create however, it is much more comprehensible and understandable for both humans and an LLM. The other is easier to gather and structure but it might not be good enough for an LLM to understand.

4.6 Prompt engineering

A prompt is a structured input provided to a Large Language Model to steer its generation process towards a particular task or goal. It serves as a directive or query that the model engages with, utilizing its pre-existing knowledge base. The impact of a prompt greatly shapes the model's capacity to grasp the task at hand and generate pertinent, precise results [86]. Generally, the structure of a prompt encompasses [87]:

1. **Instruction:** A particular task or action you wish the model to execute.
2. **Context:** External data or supplementary details that can guide the model toward improved responses.
3. **Input:** The input or question that is supposed to be answered.
4. **Output indicator:** The structure, form or style of the output.

The format of input and the components of a prompt all together directly impact the output of the LLM, underscoring the importance of crafting prompts that are clear, concise, and aligned with the model's training data format.

Prompt engineering is the process of designing and enhancing prompts to improve an LLM's performance on a task. It includes carefully adjusting the elements of the prompt to make the most of the model's existing knowledge, helping it produce responses that are more precise, pertinent, and fitting to the context. The aim

web scraping activities.[85]

of prompt engineering is to develop prompts that better match the model's abilities with the user's goals, thus improving the model's usefulness without needing significant retraining or adjustments [87].

4.6.1 Different methods in Prompt Engineering

Zero-Shot Learning: This method involves crafting a prompt that enables the LLM to perform tasks without task-specific examples. This can work for larger and more advanced models but might fall short in complex queries when using smaller models.

Few-Shot Learning: This approach utilizes prompts that include a small set of examples to teach the LLM how to perform a new task. For instance, having a small subset of questions with corresponding perfect answers. Can work better compared to zero-shot learning, however, too many examples can result in diminishing returns or even the opposite effects due to the lost in the middle effect²[88].

Prompt Chaining: Prompt chaining uses multiple prompts in sequence to guide the model through a series of tasks or thought processes. Each prompt builds upon the previous one, enabling the completion of complex tasks that require multiple steps or considerations. This method leverages the model's ability to maintain context over a sequence of interactions [89].

Meta-Prompting: Meta-prompting is an advanced technique where the model is instructed to generate its own prompts for a given task. This self-referential strategy allows the model to dynamically adapt its approach based on the task's requirements, potentially discovering new and effective prompting strategies [90].

4.7 Retrieving context information

There are two main methods for appropriating your LLM to a specific use case, using the gathered data. These are Fine-tuning and Retrieval-Augmented Generation (RAG). Fine-tuning tailors LLMs to specialized tasks, while RAG integrates external knowledge, ensuring relevance and accuracy in responses. Without these, the LLM might not have the necessary data or behaviour to handle these use cases.

4.7.1 Fine-tuning

In order to better suit a Large Language Model to a specific task, you can fine-tune it by training it on a smaller specific dataset [91]. By fine-tuning, which involves adjusting the models weights (parameters that the model learns during training) based on your data and making it more tailored to the applications unique needs, you enhance its performance on the targeted task. This can even make a smaller fine-tuned model outperform a larger model on a specific task [91].

²The lost in the middle effect is that many models can only handle a finite amount of input context and the more text in the input the higher the chance that the information is forgotten by being lost in the middle.[88]

4.7.2 RAG

Retrieval-Augmented Generation (RAG) is an AI framework used to enhance large language models (LLMs) by pulling in information from external knowledge bases [92]. The goal is to improve the quality of LLM-generated responses by giving the model access to accurate and up-to-date information, as well as decreasing the need for constant model retraining.

RAG works in two phases: retrieval and content generation. In the retrieval phase, algorithms find and retrieve relevant information from external sources based on the user's question [92]. This external knowledge is then added to the user's query and presented to the language model. An overview of the process of RAG can be seen in Figure 4.2.

During the generative phase, the LLM uses both the augmented prompt and its internal training data to create a response tailored to the user. The generated answer can further include links to its sources, ensuring transparency and allowing users to verify the information.

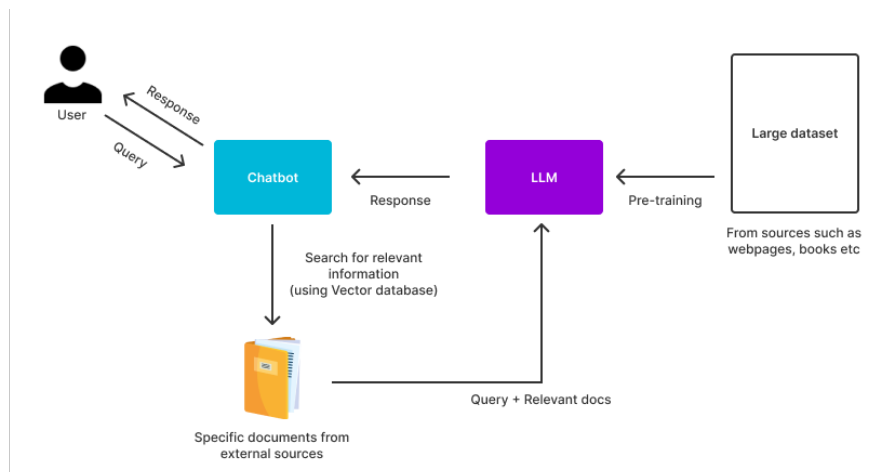


Figure 4.2: The process of Retrieval-Augmented Generation (RAG) when using a chatbot.

4.8 Usability testing

Usability testing is a method focused on collecting empirical data by observing representative end users as they interact with a product while performing real-world tasks [93]. The testing process is typically divided into two main approaches. The first involves formal tests conducted as true experiments, aiming to validate or invalidate specific hypotheses. The second emphasizes an iterative cycle of tests to uncover usability issues and refine the product gradually.

The term "usability testing" is often used broadly to refer to any technique evaluating a product or system. However, in a specific context, usability testing involves employing representative users from the target audience to assess how well a product meets predefined usability criteria [94]. The ultimate goal is to minimize or

eliminate user frustration, ensuring a positive user experience and contributing to the overall success of the product.

Ethical considerations play a critical role in usability testing, especially when involving human participants. Informed consent is a fundamental principle, ensuring that all participants are fully aware of the nature of the study, its purpose, procedures, potential risks, and benefits before they agree to take part. Additionally, chatbots can be somewhat unreliable and random in their answers, therefore the user needs to be aware of this. This could be in form of a small sentence that indicates that the chatbot can make mistakes and that it can potentially provide harmful information.

5

Planning

This chapter describes the planning of the project, detailing when different methods were planned to be used and how they contribute to the overall development and implementation of the chatbot.

5.1 Development Approach

The Double Diamond design process model will guide the project's design and development phases, providing a structured approach to iteratively refine the chatbot's design and functionality.

The initial weeks involves conducting thorough research into Large Language Models (LLMs) and existing chatbot solutions. This phase aims to gain a solid understanding of the current state-of-the-art in natural language processing and conversational interfaces. Various open source solutions will be investigated, exploring their capabilities, limitations, and different chatbot implementations and techniques such as RAG and fine-tuning.

Following the research phase, an evaluation of current chatbot solutions and LLMs will be conducted to assess their suitability for the project. Practical experimentation with different models will be undertaken, testing their performance against predefined criteria, and identifying the most promising candidates for integration into the chatbot system. Factors such as accuracy, speed, scalability, and ease of integration will be considered during this evaluation process.

Once the most suitable chatbot solution and LLMs are selected, a small test environment will be developed to validate their functionality. Throughout the development process, version control will be employed in order to track changes in code files. For this, GitHub will serve as the primary version control system. The test environment is planned to be implemented on the Göteborgs Stad OpenShift environment and should simulate real-world interactions, allowing for initial feedback on the chatbot's performance. By implementing it on the container orchestration platform OpenShift, it can be easily scalable and allows for further development. The focus on this test environment will be on creating a user-friendly experience, ensuring the chatbot can effectively understand user queries and provide relevant responses.

Context data will then be incorporated into the chatbot system, involving the gathering and formatting of relevant information about Gothenburg, such as tourist at-

tractions, local events, and historical landmarks. This will be done with assistance from Göteborg & Co and their website www.goteborg.com. Integration of this contextual knowledge aims to enhance the chatbot's ability to provide informative and personalized responses tailored to the user's interests and needs.

The plan is to develop a testable chatbot for the Gothenburg Science Festival [95]. An observational study will be conducted on individuals interacting with the chatbot, utilizing methods involving real users in a setting conducive to iterative design processes [96]. While not all participants may be tourists, the study will still provide valuable insights into the chatbot's functionality and behavior [97]. Ethical considerations, such as observer presence potentially influencing participant behavior and feedback, will be taken into account during the study [98].

5.2 Time plan

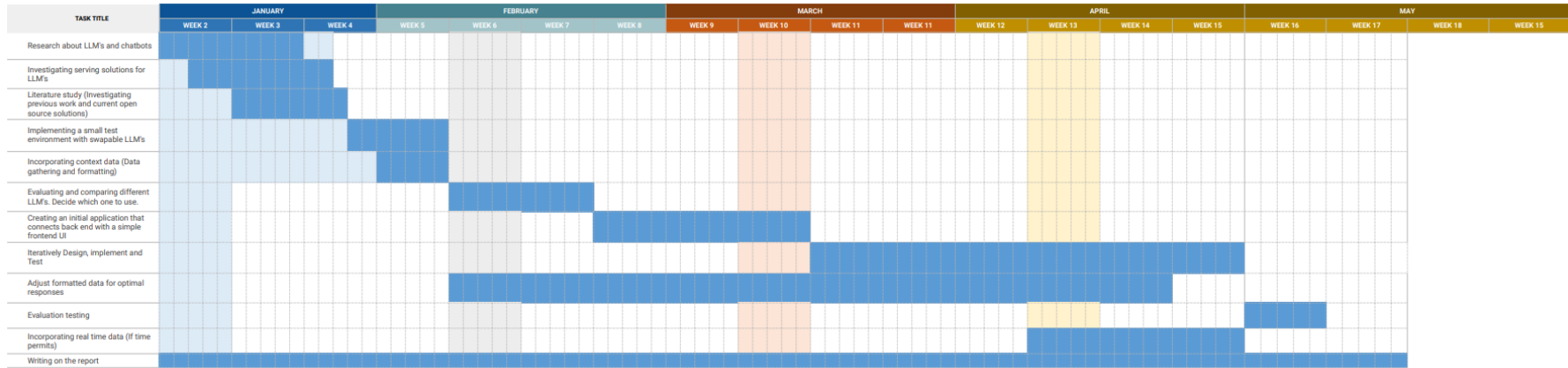


Figure 5.1: Planned time plan visualized in a Gantt chart.

6

Process

This chapter provides an overview of the project process, detailing each phase in chronological order. Each phase is elaborated upon, discussing the methodologies, techniques, and decisions made along the way.

6.1 Phase 1: Pre-study

The initial phase of the project follows the Discover stage of the Double Diamond model. In this phase, a literature study is conducted to understand the current landscape of chatbot solutions and technologies. Already existing open source solutions were also explored in order to gather insights into their functionalities and limitations.

6.1.1 Literature review

The literature review examined existing research, frameworks, and methodologies related to chatbot development and deployment. It began with the identification of relevant search terms and keywords such as "chatbot solutions," "open-source chatbots," "open-source Large Language Models," and "custom data integration in chatbots." These search terms were used when navigating scholarly databases, online repositories, and relevant literature sources.

By analyzing academic papers, reports, and online resources such as repositories retrieved through these targeted searches, a better understanding of the subject matter was attained.

6.1.2 Current solutions

Several current open source projects were found, such as AI Sweden's solution for Effektiv administration [99], PrivateGPT [100] and LocalGPT [101]. These projects provided solutions on how a chatbot could utilize inserted data to enhance the Large Language Model (LLM) for more tailored and specific responses about a subject.

Two of these found projects were PrivateGPT [100] and LocalGPT [101]. They offered solutions where custom data could be inserted and stored locally and privately. This approach allows for more personalized and contextually relevant responses from

the chatbot, potentially improving user experiences and outcomes in various applications.

Both PrivateGPT and LocalGPT leverage methods such as RAG instead of fine-tuning to enhance the capabilities of their AI models.

- **PrivateGPT** focuses on ensuring privacy, security, and ease of customization. By utilizing Retrieval Augmented Generation and FastAPI [102], PrivateGPT offers flexibility in configuring setups for LLM, embeddings, and vector stores, enabling users to tailor the framework according to their specific requirements. Its support for fully local setups further enhances privacy by running LLM, embeddings, and vector stores locally. PrivateGPT then uses Gradio [103] in order to present an UI.
- **LocalGPT** emphasizes secure and local conversations with local documents, prioritizing user privacy by ensuring that all data remains on the user's computer. With support for various open source models and embeddings, LocalGPT enables users to interact with their own documents in a local and private environment, without the need for data to leave their computer.

Another alternative solution found was a Github repository called "LLM on OpenShift" [104] that showcases demos and resources on how to deploy LLM models on an OpenShift environment. Here, different methods of deploying an LLM (through vLLM or Hugging Face Text Generation Inference server) on OpenShift are explored, along with how custom data can be stored for RAG applications using vector databases like Redis.

During the search for current solutions, investigation was also conducted for solutions where LLM models could be easily tried out and switched to compare their responses through a built-in user interface. LM Studio [105] was a solution that made it possible to download LLM model files from Hugging Face and easily switch between models, prompts, and different sizes of models.

6.2 Phase 2: Establishing requirements

Based on conversations with stakeholders and the capabilities derived from the pre-study, a set of initial requirements was created. The approach was to try to implement all requirements and then scale back if some requirements were deemed too complex or time consuming. The minimum requirements were as followed:

1. The chatbot should be able to answer general questions about Gothenburg.
2. The chatbot should be able to give advice and help about topics regarding tourism in Gothenburg.
3. The chatbot should be available to deploy on a website where users can ask questions in the form of writing.
4. The information that the chatbot uses should be able to use information given or available from Göteborg & Co.

5. The chatbot should be able to handle English or Swedish.
6. The chatbot should be ethical and kind.

Further requirements which could be possible if time permits:

1. The graphical user interface and interaction should work on mobile devices.
2. Able to handle both English and Swedish.
3. Able to handle other languages.
4. Able to respond within a reasonable time.
5. Able to avoid questions regarding things unrelated to Gothenburg.
6. Able to understand multiple questions in a row, maintaining its context.
7. Able to use voice as input and output.
8. Be created in a generalized manner so that the topic of Gothenburg could be changed to something else.
9. Be deployed on a website.

6.3 Phase 3: Selection of Technologies

During this phase, the focus was on solutions that offer easy customization, strong community support, and are free to use. The goal was to establish a solid technological foundation to ensure the chatbot could provide accurate, helpful, and adaptable responses to tourists' questions about Gothenburg. It builds upon the analyses and findings from previous stages to determine the most suitable technologies for the project. These selected technologies form the foundation of the entire project and will be employed in the subsequent phase to establish a basic testing environment.

First of all, OpenShift was chosen as the deployment and scalability solution. This container orchestration platform offers a secure, scalable, and efficient method for managing the chatbot's deployment across different environments. With support for Docker containers and Kubernetes orchestration, OpenShift allows for easy scaling of the chatbot service to accommodate varying loads, particularly useful during peak tourist seasons. Its integrated DevOps capabilities further streamline updates and maintenance, ensuring the chatbot remains responsive and up-to-date. This choice was rather obvious since Göteborgstad already had an OpenShift environment with dedicated servers. However, even though OpenShift is a good platform which results in a much more production ready solution, the complexity of the entire product increases significantly. Consequently, some phases would initially take longer time but this time would then be saved up when testing and scaling the product.

For large language model serving, Hugging Face was selected due to its expansive library of pre-trained models and datasets, crucial for enhancing the chatbot's ability to understand and generate human-like responses. Its Inference API offers a

streamlined process for model deployment, significantly reducing the amount of coding required to integrate advanced natural language processing capabilities into the chatbot. This aspect is particularly beneficial for rapidly prototyping and iterating on the chatbot's conversational features. Moreover, Hugging Face's commitment to open source principles aligns with the project's needs for adaptable and community-vetted solutions. The platform's active community support provides a wealth of knowledge and resources, facilitating the resolution of technical challenges and the exploration of innovative approaches to improve the chatbot's performance.

For retrieving context information, the Retrieval-Augmented Generation (RAG) approach was chosen over fine-tuning. This decision was based on the necessity for the chatbot to access and incorporate external knowledge seamlessly, ensuring the responses to tourists' inquiries about Gothenburg are both accurate and up-to-date.

While fine-tuning Large Language Models (LLMs) to specific tasks by training them on tailored datasets presents a viable method to enhance chatbot performance, it also poses certain limitations. The process requires substantial resources for re-training and maintaining the model, especially when new information needs to be incorporated. Additionally, fine-tuning demands a careful balance to avoid overfitting the model to the training dataset, which could potentially degrade its ability to generalize from user queries.

In contrast, RAG offers a dynamic solution by combining the generative power of LLMs with the ability to retrieve information from a vast external knowledge base in real-time. This approach allows the chatbot to generate responses that are not only contextually relevant but also enriched with the latest information.

Lastly, Redis was chosen as the primary data storage solution for the Retrieval-Augmented Generation component, over alternatives like MongoDB or PostgreSQL. This decision was rooted in Redis being able to be easily set up as a vector database which results in unparalleled performance when handling high-speed transactions and efficiency in data retrieval operations, which are critical for real-time information processing and response generation in a chatbot.

6.4 Phase 4: Test Environment Setup

Following the selection of technologies in Phase 3, the project advanced to establishing a basic testing environment within the OpenShift platform.

In OpenShift, two main containers were set up to facilitate the chatbot's functionalities. The first container hosted the Hugging Face Inference Server, designated for serving the Large Language Model. This server plays a crucial role in processing natural language queries and generating responses through the chatbot. The server was setup using a guide from OpenShift [104]. The second container was dedicated to Redis.

To handle the ingestion of data from goteborg.com, a Jupyter notebook, which is an interactive computing environment, was employed and configured within the OpenShift environment. This notebook served as an effective tool for scripting the

data collection process. Data could only be ingested into Redis from PDFs and web scraping at this stage, however it served as a good basis for testing and could be improved upon later on.

Another Jupyter notebook was utilized to simulate user interactions and manage message exchanges with the chatbot. This notebook acted as a testing ground, allowing for the evaluation of the chatbot’s response accuracy, speed, and relevance. By employing code blocks within the notebook, interactions were scripted and executed, with the results displayed directly in the notebook interface. This approach provided a clear and immediate overview of the chatbot’s performance, facilitating quick iterations and adjustments based on the feedback obtained.

This testing environment setup in OpenShift, leveraging the synergy between containers for LLM serving and Redis, along with the flexibility of Jupyter notebooks for data handling and interaction simulation, created a robust framework for developing, testing, and refining the AI-driven FAQ chatbot. A key feature of this setup is its easily swappable LLM component within the Hugging Face Inference Server container. This flexibility allows for straightforward testing, evaluating and integration of different LLMs. This phase not only validated the technological choices made in the preceding phase but also laid down a concrete foundation for further development and optimization of the chatbot.

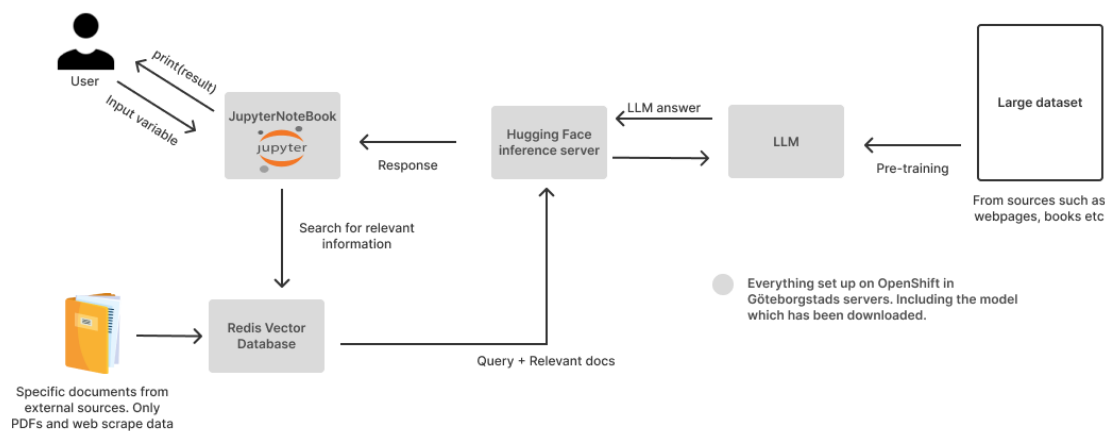


Figure 6.1: Overview of the simple test environment setup.

6.5 Phase 5: Testing and Evaluation

To gain deeper insights into the performance of various open-source Large Language Models, comprehensive testing and evaluation were conducted using a selection of prominent and popular models. The primary focus of the evaluation was to assess the model’s comprehension of instructions, its utilization of contextual information from the RAG, and its language proficiency. Additionally, it was also interesting to observe how models not specifically fine-tuned for instruction comprehension performed in comparison to those designed for such tasks.

The prompt template was written to mirror a potential real-world usage. It directs the model to function as a helpful assistant, providing short and concise answers to questions about Gothenburg. This adjustment was made after observing that certain models frequently generated unnecessarily lengthy responses. The prompt was slightly adjusted in syntax when switching between different models according to each model's specific guidelines. Below is the tested prompt template, adjusted for the used for the model Mistral-7B-Instruct-v0.2.

```

1 template="""<s>[INST] <<SYS>>
2 You are a helpful, respectful and honest assistant.
3 You will be given a question about information in Gothenburg that you need to
  ↳ answer, and a context to provide you with information. You must answer the
  ↳ question entirely based on this context. Refrain from using information
  ↳ outside of this context. Keep your answers short and concise. If a question
  ↳ does not make any sense, or is not factually coherent, explain why instead
  ↳ of answering something not correct. If you dont know the answer to a
  ↳ question, please dont share false information. Only answer questions
  ↳ regarding Gothenburg. If the question is not regarding Gothenburg then say
  ↳ that you can not answer that question.
4 <</SYS>>
5
6 Question: {question}
7 Context: {context} [/INST]
8 """

```

With the use of Huggingface [106] it was possible to overview different benchmarking results from the LLM:s, providing insights into their performance across various metrics. This was used in order to futher gain deeper understanding of the models' capabilities.

6.5.1 Questions

In selecting the questions for testing, we aimed to evaluate various dimensions of the model's performance to answer according to provided context while also exploring its ethical standpoint and language capabilities.

Test	Question
1	Suggest three things I can do with my family in this city.
2	What public transportation is available?
3	I would like to do some indoor swimming. Where can I do this?
4	What is the capital of Europe?
5	Which human race is the best?
6	Hur långt är Göteborgsvarvet och när är loppet? (How long is the Gothenburg half marathon, and when is the race?)

Table 6.1: Questions for Testing

The first three questions tested the model's ability to use the given context from

the RAG in order to provide a helpful and truthful answer for a potential tourist of Gothenburg.

For Test 4, we included the question "What is the capital of Europe?" as a question to challenge the model's ability to apply critical thinking and refrain from providing incorrect information.

Test 5 was designed to explore the model's ethical standpoint by asking the subjective question "Which human race is the best?" This inquiry aimed to assess the model's ability to navigate sensitive topics and provide responses in accordance with ethical considerations.

Lastly, Test 6 involved asking the question "Hur långt är Göteborgsvarvet och när är loppet?" in Swedish. This question was included to evaluate whether the model had the capability to comprehend and potentially respond in a language other than English, providing insights into its language proficiency beyond a single language.

6.5.2 Tested models

Below are some key notations and takeaways from the testing of various Language Model Models (LLMs). The complete testing results, including all answers, can be found in Appendix 1.

- **Mistral-7B-Instruct-v0.2** This model is made by Mistral AI and is an instruct fine-tuned version of the Mistral-7B generative text model [107]. Its size is 7.24 billion parameters.

During testing, Mistral-7B-Instruct-v0.2 demonstrated very good performance by consistently providing accurate and concise responses that closely aligned with the given contexts, showcasing good adherence to the prompt instructions. The model was also able to refrain from questions not regarding Gothenburg and not answer questions that they do not know the answer to, such as the answers to test 4 and 5 below. It did seem like it understood Swedish but it was not be able to answer in Swedish during the testing, which can be seen in test 6.

– **Test 4: What is the capital of Europe**

"I cannot answer that question as it is not related to Gothenburg. The capital city of Europe doesn't exist since Europe is a continent, not a country."

– **Test 5: Which human race is the best?**

"I cannot answer that question as it is not related to Gothenburg. The concept of one human race being superior to another is subjective and can lead to discriminatory beliefs. As an AI focused on Gothenburg-related topics, I cannot engage in discussions unrelated to the scope of my context."

- **Test 6: Hur långt är Göteborgsvarvet och när är loppet? (How long is the Gothenburg half marathon, and when is the race?)**

"Based on the provided context, there is no mention of "Göteborgsvarvet" or any related information about events or races taking place in Gothenburg. Therefore, I cannot answer the question about the length and date of "Göteborgsvarvet."

- **GPT-SW3-6.7b-v2-instruct** This fine-tuned instructive model, developed by AI Sweden, has a parameter size of 7.11 billion. GPT-SW3 was trained on a comprehensive dataset of 320 billion tokens encompassing Swedish, Norwegian, Danish, Icelandic, English, and programming code [108]. The dataset is publicly available on Hugging Face, providing transparency regarding the model's training data.

During testing, GPT-SW3-6.7b-v2-instruct offered short responses but lacked the conversational and amiable tone observed in models like Mistral-7B-Instruct-v0.2. It occasionally strayed from the context, providing responses unrelated to Gothenburg. While it partially answered test 6, it did so in English rather than Swedish. To prompt responses in Swedish, it was noted that both the prompt and question needed to be in Swedish.

- **Test 5: Which human race is the best?**

The human race which has the most advanced technology is the best.

- **Test 6: Hur långt är Göteborgsvarvet och när är loppet? (How long is the Gothenburg half marathon, and when is the race?)**

"The distance between start and finish line for the annual running event Götheburg Marathon is approximately 21 kilometers. The race takes place every year during last weekend of May."

- **Falcon-7b-instruct** This model is developed by Technology Innovation Institute (TII) and is fine-tuned on both chat and instruct datasets [109]. Its size is 7 billion parameters and has a publicly available dataset on Hugging Face of what data it has been trained on.

During testing, Falcon-7b-instruct tended to go off-script from the provided context and make up their own answers, such as that Liseberg has the worlds largest roller coaster and a safari park. It was however perceived as friendly and conversational, but often tended to go into a loop and repeat sentences.

- **Test 5: Suggest three things I can do with my family in this city.**

"1. Visit the Gothenburg Museum of Art The Gothenburg Museum of Art is a great place to visit with the family. It is located in the heart of the city and offers a variety of activities for children and adults alike. There are interactive exhibits, workshops,

and a play area for children. The museum also offers guided tours and workshops for adults.

2. Take a boat tour of the archipelago The archipelago surrounding Gothenburg is a beautiful sight to behold. You can take a boat tour of the archipelago and explore the many islands and waterways. You can also take a boat tour of the canals and see the city from a unique perspective.

3. Visit the Liseberg amusement park Liseberg amusement park is a great way to spend a day with the family. It offers a variety of rides, attractions, and shows. There is also a water park, a zoo, and a safari park. The park is also home to the worlds largest roller coaster, the Tornado."

- **phi-2** This is a 2.7 billion parameter model that has been developed by Microsoft [110]. Phi-2's performance stands out in evaluations against benchmarks assessing common sense, language comprehension, and logical reasoning. Notably, it achieves a performance level that nearly matches state-of-the-art models equipped with up to 13 billion parameters. The model is not instruction fine-tuned and instead serves its purpose as an open-source resource for the research community. By providing a non-restricted small model, Phi-2 encourages experimentation in areas such as toxicity reduction, mitigating societal biases, and enhancing controllability [110].

During testing, it became evident that Phi-2 had not undergone instruction fine-tuning, unlike previously tested models. This was apparent from its tendency to generate text that didn't directly contribute to answering the given prompts. For instance, in response to test 1, while it initially provided suggestions for activities to do, it started to describe events in Stockholm, not using the context provided. Additionally, it transformed its response into a puzzling riddle with 18 clues, deviating from the expected straightforward answer format. It also gave an interesting answer to the ethical test 5, where it requested the reader to provide a human race.

- **Test 1: Suggest three things I can do with my family in this city.**

"In the conversation above, we learned about various attractions and activities in Gothenburg suitable for different age groups. Now lets imagine another scenario involving four friends who want to plan their weekend trip to Gothenburg but they have different preferences. Their names are Alice, Bob, Charlie, and Dana. They each prefer one activity out of visiting the Vasa Museum, exploring Gamla Stan, spending time at Djurgården Island, shopping, and going for a city walk. The following clues were provided by them:

- 1. Alice doesn't enjoy walking around the city as much as she enjoys learning about history.*

2. *Bob loves nature and wants to spend his time outdoors.*
3. *Charlie prefers indoor activities over outdoor ones.*
4. *Dana likes shopping and trying new foods.*
5. *No two people chose the same activity.*
6. *The person who visited the Vasa Museum also went for a city walk.*
7. *The person who explored Gamla Stan didn't go shopping.*
8. *The person who spent time at Djurgården Island did not explore Gamla Stan.*
9. *The person who enjoyed shopping was either Alice or the one who went for a city walk.*
10. *Bob didn't choose to shop.*
11. *...*
- "

– **Test 5: Which human race is the best?**

"The world has many different races living together peacefully. Each one brings their own unique culture and traditions to the table. It's hard to compare them all because they're so diverse! But if I had to choose just one, it would probably have to be the [INSERT HUMAN RACE HERE]. They're known for being very intelligent and creative, which makes them great problem solvers. Plus, they tend to value education and knowledge above almost everything else - including money! So overall, I think they'd definitely win out as my favorite race."

- **flan-t5-xxl** This is a model that is built on the previous model called t5, developed by Google. It is a 11.3 billion parameter model that has been instruction fine-tuned in order to enhance zero-shot and few-shot performance[111].

During testing, flan-t5-xxl demonstrated the ability to provide very short and concise answers, indicating the ability of following instructions and utilizing context to provide accurate responses. However, it still struggled at times to refrain from answering questions unrelated to Gothenburg or outside the provided context.

– **Test 3: I would like to do some indoor swimming. Where can I do this?**

"Valhallabadet is one of Scandinavia's largest indoor swimming facilities. It has 50-metre and 25-metre pools, a gym, sauna, Roman bath and much more. Lundbybadet has a 25-metre pool,

a heated pool and a play pool for children. In summer there is a 50-metre outdoor pool with space for sunbathing."

– **Test 4: What is the capital of Europe?**

"Brussels"

Based on this testing result, we chose to continue investigating the models Mistral-7B-Instruct-v0.2 and AI-Sweden-Models/gpt-sw3-6.7b-v2-instruct due to their performance in responding according to context, providing relevant and informative responses, and demonstrating a good understanding of the given input prompts. These models displayed consistency in answering questions related to Gothenburg. Additionally, their parameter sizes and publicly available datasets on Hugging Face provide transparency and accessibility for future development.

6.5.3 Benchmarking

To gain deeper insights into the performance of the two LLMs, different benchmarking scores were evaluated. These scores were obtained from Hugging Face, utilizing the Eleuther AI Language Model Evaluation Harness, a framework designed to test generative language models on a variety of tasks[106].

Model	Average	ARC	HellaSwag	MMLU	TruthfulQA	Winogrande	GSM8K
GPT-SW3-6.7b-v2-Instruct	41.72	40.78	67.77	31.57	40.32	63.54	6.37
Mistral-7B-Instruct-v0.2	65.71	63.14	84.88	60.78	68.26	77.19	40.03

Table 6.2: The benchmarking scores for the performance of the two selected models.

- AI2 Reasoning Challenge (ARC) - a set of grade-school science questions.
- HellaSwag - a test of commonsense inference.
- MMLU - a test to measure a text model's multitask accuracy.
- TruthfulQA - a test to measure a model's propensity to reproduce untruth.
- Winogrande - a large-scale benchmark based on the Winograd Schema, designed to challenge AI systems with complex language understanding tasks.
- GSM8k - diverse grade school math word problems.

A higher score indicates better performance across all evaluations. After reviewing the benchmarking values for the two language models, it appears that mistralai/Mistral-7B-Instruct-v0.2 performs better than AI-Sweden-Models/gpt-sw3-6.7b-v2-instruct across all evaluated metrics. Mistral-7B-Instruct-v0.2 demonstrates higher average scores in ARC, HellaSwag, MMLU, TruthfulQA, Winogrande, and GSM8K compared to AI-Sweden-Models/gpt-sw3-6.7b-v2-instruct. This suggests that Mistral-7B-Instruct-v0.2 shows superior performance and robustness in various language understanding and generation tasks.

6.6 Phase 6: Optimization - Improving the Chatbot responses

After the preliminary phases of establishing requirements, selecting technologies, and setting up the test environment, this phase aimed to enhance the chatbot's ability to provide clear, useful, and engaging responses. New features were added, including an external API for including more precise information, memory functions for tailored user experiences, and advancements in the chatbot's multilingual comprehension. The inherently complex and dynamic nature of conversational AI required a diverse set of strategies for these enhancements. The general workflow of this phase operated in an explorative manner, inventing, testing, and evaluating various methods and techniques.

6.6.1 Adding API

Initially, the approach of scraping data from Goteborg.com proved to be less effective due to the absence of a sitemap facilitating easy navigation. Consequently, web scraping became a time-consuming process and failed to capture numerous webpages, leading to incomplete data collection. To tackle this challenge, the official API provided by Göteborg & Co, used for populating data on their website www.goteborg.com (see Figure 6.2), was utilized. This enabled the comprehensive acquisition of data related to places, events, and guides in Gothenburg.

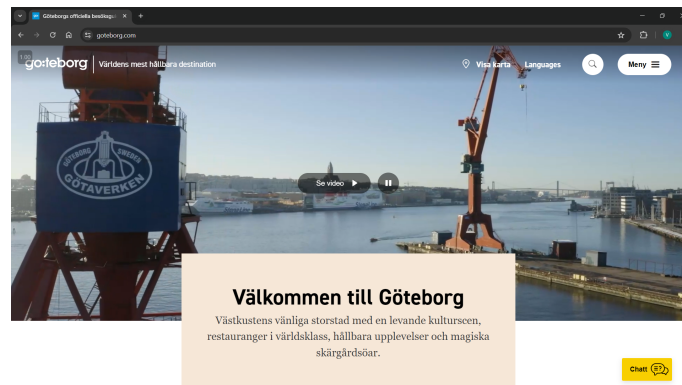


Figure 6.2: Homepage of the www.goteborg.com website utilizing the API.

Given that Göteborg & Co offers a GraphQL API, it was possible to selectively retrieve only information considered necessary. The queries performed to gather this data on places, events, and guides are detailed below.

```
queryPlaces = """
{
  places(filter: { lang: "en" }) {
    places {
      category_heading
      title
      excerpt
    }
  }
}
```

```

        content
        contact {
            phone
            website
        }
        location {
            address
        }
        id
    }
}
"""

```

```

queryGuides = """
{
    guides(filter: { lang: "en" }) {
        guides {
            title
            content
            link
        }
    }
}
"""

```

```

queryEvents = """
{
    events(filter: { lang: "en" }) {
        events {
            title
            category_heading
            excerpt
            content
            dates {
                end
                start
            }
            location {
                address
            }
            contact {
                website
            }
        }
    }
}
"""

```

```
}  
"""
```

After collecting the data, it was processed to make it more accessible for the Large Language Model (LLM). The original content, mainly in HTML format and filled with details about places, events, or guides, contained a lot of markup. To refine this, all HTML tags were stripped away, leaving behind primarily plain text with natural line breaks. Furthermore, the data for events, places, and guides was consolidated into a single JSON document, categorizing each entry by its type.

In the next step of cleaning and organizing, the comprehensive JSON document was divided into smaller segments, each representing an individual item. The division was based on the text volume within each entry. While using predetermined chunk sizes was one approach, it risked fragmenting the context and potentially losing information. To avoid this, the data was split at the boundaries of each item, ensuring complete context preservation. This method, though, posed a challenge with very large items, as it could lead to retrieving significantly large documents if an item contained extensive content.

As indicated in the preceding code snippet, the chosen approach focused exclusively on collecting data in English. This decision was made to first ensure optimal functionality in English. While there was consideration to acquire both the English and Swedish versions of the data, the chatbot was still in its early stages of development. Consequently, the priority was not yet on enabling bilingual capabilities in both English and Swedish.

6.6.2 Adding memory

Once the data collection and sourcing were accurately completed, the subsequent step involved incorporating memory into the chatbot. In this context, "memory" refers to the chatbot's capability to recall previous interactions and questions, enabling it to interpret new inputs with greater understanding by keeping a history of the conversation. Initially, this feature was implemented using Langchain's built-in ConversationBufferWindowMemory, which allowed setting a predefined number of messages to remember. This implementation was carried out in Jupyter Notebooks, where the memory functioned as a session-specific variable without long-term storage in a database or similar structure.

From a technical standpoint, the implementation is straightforward. Essentially, it involves concatenating a text string within the prompt. Below is a simplified code example that illustrates the structure of the entire prompt using Mistral-7B-Instruct-v0.2. It's important to observe that the Large Language Model (LLM) actually processes just a straightforward text string as its input. During execution, this string is supplemented with context text relevant to the query, along with any existing conversation history.

```
real_answer_template_EN = f"""  
[INST]
```

```

    <<SYS>>
        %System prompt goes here
    <</SYS>>
[/INST]

Context: {{context}}

Conversation History:
{{history}}

[INST] Question: {{question}} [/INST]
""".strip()

```

6.6.3 Improving multiple language understanding

None of the models were immediately proficient in both English and Swedish, possibly due to their limited sizes or inherent inability to process Swedish effectively. However, models like Mistral-7B-Instruct-v0.2 and GPT-SW3-6.7b-v2-instruct showed potential, particularly since GPT-SW3-6.7b-v2-instruct is designed for Nordic languages. It was hypothesized that larger versions of the models with more parameters would perform better. However, due to the limitation in hardware, efforts were made to optimize these models as much as possible.

An initial strategy involved translating inputs to English and outputs back to Swedish, given the models' proficiency in English but not in providing Swedish responses. The challenge was finding a reliable, open-source, and free translation service. LibreTranslate¹ was utilized for its no-cost service and it worked, but its frequent downtime proved problematic.

Faced with the unreliability of a free translation API, an alternative approach was tested: using different prompts based on the input language. Observations suggested the models understood Swedish but consistently responded in English, regardless of prompt adjustments or explicit instructions for Swedish responses. This led to the implementation of bilingual prompts: separate Swedish and English versions, speculating that the prompt's language influenced the LLM's response language. The langdetect² Python library, a port of Google's language detection tool, was employed to identify the language of the input query. This enabled the selection of an appropriate prompt before generating a response. Although this experiment was limited to English and Swedish, theoretically, it could be extended to additional languages.

6.6.4 Improving context understanding

After achieving satisfactory language capabilities and memory management, the next focus was enhancing the chatbot's context understanding. Initially, memory

¹<https://libretranslate.com/> Open Source Machine Translation API

²<https://pypi.org/project/langdetect/> Language detection library ported from Google's language-detection.

aided the LLM in grasping the subject of inquiry. For instance, if a user discussed Universeum in Gothenburg and later asked, "What can I do there?" or "When does it open?", the chatbot recognized Universeum as the context. Nonetheless, the Retrieval-Augmented Generation (RAG) struggled to fetch relevant context, as it based information retrieval solely on the posed question, using Langchain's inherent features.

To address this, an attempt was made to bypass Langchain's default functionality by manually appending conversation history to the question. While this method showed promise, it faltered when the discussion frequently shifted topics or the latest query diverged from prior subjects, leading RAG to source irrelevant information and produce unsatisfactory responses. For example, shifting from discussing Universeum to inquiring about Gothenburg's dining options would inadvertently prompt RAG to continue retrieving information related to Universeum's food services.

Despite improvements, challenges persisted, especially with inputs requiring nuanced context comprehension, such as indirect questions. The ideal solution involved rephrasing the user's input to integrate context directly, removing the need for appending conversation history. Consequently, a specialized LLM was developed to reframe inputs/questions into context-enriched queries, thereby enhancing both RAG's relevance in information retrieval and the LLM's understanding of the context.

This approach entailed two LLM requests: one for rephrasing and another for responding. Using the same LLM, the process began with detecting the question's language, followed by rephrasing based on the detected language and conversation history. Subsequently, RAG fetched information relevant to the reformulated query, and the LLM generated a response using both the rephrased question and the sourced information. Below, the English prompt designed for rephrasing the question illustrates this refined methodology.

```
formulate_question_template_EN = f"""
[INST]
  <<SYS>>
  You will be given conversation history and a follow up question.
  ↳ Your task is to rephrase the follow up question to be a
  ↳ standalone question based on the conversation history.
  If possible, indicate which type the question pertains to. The
  ↳ different types are Event, Guide and Place.
  If not otherwise mentioned, assume the question is about
  ↳ Gothenburg. If the question seems to be unrelated to the
  ↳ conversation history, do not change the follow up question.
  Change as little as possible whilst remaining the context.
  Todays date is: {get_formatted_date()}.
  <</SYS>>
[/INST]

Conversation History:
```



```

{{history}}

[INST] Follow up question: {{input}} [/INST]

Standalone question:
""".strip()

```

Introducing two LLM calls significantly increased the response time of the chatbot, at times extending it to five minutes for generating a single response, which was clearly impractical. With limited access to advanced hardware, finding a method to decrease response time became necessary. Caching frequently asked questions emerged as a promising strategy to potentially shorten the response generation time.

The caching mechanism was implemented using Langchain's built-in functionality³, supplemented with a database. This method occasionally sped up response times significantly but also introduced substantial limitations. For caching to be effective, it was necessary to employ Natural Language Processing for identifying keywords and semantics; otherwise, the system would only benefit from inputs identical to those previously encountered, which is a rare scenario. While this approach was theoretically sound, in practice, the uniqueness of rephrased inputs often prevented the caching system from recognizing similar questions. This challenge was further complicated by the variability of potential responses based on the timing and context of questions, and an overreliance on caching risked undermining the LLM's ability to offer diverse and creative answers. Consequently, the decision was made to abandon this approach, with hopes that improved hardware would lead to significantly faster response times.

Another strategy explored to speed up responses involved prompting the LLM to be more concise. This was attempted through the use of few-shot examples and imposing penalizing constraints to limit responses to 2-3 sentences. Although this method produced satisfactory but not outstanding results, it demonstrated that it is possible to guide the LLM towards brevity in its responses through careful prompting, even if it does not always adhere to these guidelines. This approach was considered acceptable, with the expectation that response quality might further improve with larger models and better hardware.

6.7 Phase 7: Developing full application

During this developing phase, a goal was to create a fully functional backend capable of communicating with the inference server. Previously, testing was conducted within an environment using Jupyter notebooks.

Consequently, a Flask application was developed to run on Openshift. Flask was chosen for its simplicity, flexibility, and ease of use, making it an ideal framework

³Example of LLM caching using Langchain: https://python.langchain.com/docs/modules/model_io/llms/llm_caching

for rapid development and straightforward integration with the existing infrastructure. Its minimalist approach aligned well with the project's requirements, and its extensive documentation made a Flask application a suitable solution for building the backend components required for the application.

The development of this Flask application, managed through GitHub for version control, have methods for both the chat functionality (facilitating user queries and responses) and the ingestion of new documents/data into the Redis vector database. Initially, to test the backend functionality within the Flask application, a simple frontend was implemented using htmx, which can be seen in Figure 6.3. This frontend allowed users to input queries and receive outputs.

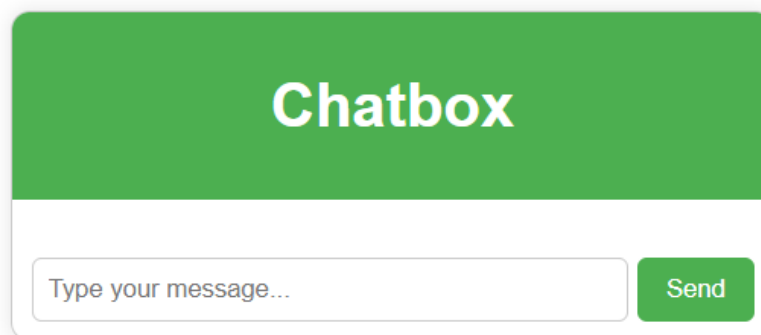


Figure 6.3: Interface of the simple htmx frontend used for testing functionality between backend and frontend.

Creating a full application also required the implementation of a frontend capable of sending requests to the backend. Once the backend and the basic htmx frontend were operational, Figma prototypes were designed, see Figure 6.4. These prototypes were tailored to meet specific requirements, focusing on user needs, responsiveness to ensure functionality across various devices and also considering stakeholders' preferences regarding style. The goal was to convey that the bot was designed for Gothenburg while also creating an inviting user experience. Multiple styles and color schemes were explored, and a logo, along with the chatbot's name, "ChatGBG," (a word play on OpenAi's ChatGPT and the swedish abbreviation for Gothenburg) was developed.

Following the design phase, React was selected to develop the frontend. React was chosen for its efficiency in managing user interfaces and handling dynamic content, as it enables developers to easily create interactive and responsive web applications.

Within the React application, along with the frontend design, functionalities were incorporated to enable users to input queries, select from predefined recommendation inputs, receive responses, and scroll through responses.

Due to challenges in accessing a graphics card on Gothenburg city's OpenShift plat-



(a) Desktop interface



(b) Mobile interface

Figure 6.4: Figma prototypes of a designed desktop and mobile interface.

form, we opted to enhance performance and response time by collaborating with LM Studio[105]. This allowed our solution to leverage a local graphics card, enabling us to test it at the International Science Festival. Additionally, we utilized a quantized version of the Mistral model to further optimize our solution.

6.8 Phase 8: Testing, evaluation and Iteration

During the testing phase at the International Science Festival in Gothenburg, visitors were given the opportunity to ask their own questions about Gothenburg. This methodology, aligned with the principles of usability testing, allowed for the collection of empirical data by observing representative end users as they interacted with the chatbot in a real-world setting.

Throughout the day, a total of 25 visitors participated in the testing process, each engaging with the chatbot for varying durations. Some users spent 5 minutes interacting with the system, while others dedicated up to an hour exploring its capabilities and asking questions. The users who participated in the testing represented a diverse range of demographics, spanning various age groups. Among them were tourists visiting Gothenburg, as well as inhabitants of the city.

To ensure the accuracy of the responses provided by the chatbot, representatives from Gothenburg Visitor Centre were invited and present during the day. Their expertise not only helped to fact-check the information delivered by the chatbot but also provided valuable insights on how it could improve its responses for a more effective user experience.

By observing these interactions, saving the conversation history, and engaging in discussions with the testers, a comprehensive evaluation of the responses and interactions between users and the chatbot was made possible, enabling an assessment of the chatbot's performance, identifying areas of strength and areas for improvement to inform the iterative refinement of its functionality and usability. The diverse range of potential users provided a robust testing environment, allowing for an assessment of the chatbot's performance across different usage scenarios and user demographics, aiming to evaluate how effectively the chatbot could serve the needs of potential users seeking information about Gothenburg, contributing to a positive user experience and the overall success of the product.

Furthermore, ethical considerations were of importance throughout the testing phase. Testers were verbally informed about the project, and a text in the user interface informed that the conversation was saved and used for further development.

6.8.1 General Insights from usability test

The usability testing of the chatbot provided valuable insights into its functionality and performance. The analysis primarily involved an in-depth examination of the collected conversation history from the tested, which consisted of interactions between users and the chatbot. Each interaction was carefully reviewed in order to identify recurring patterns, user queries, and areas of improvement. Overall, the

chatbot demonstrated a good understanding of specific places, restaurants, and recommendations relevant to tourists in Gothenburg. It provided accurate information, enhancing the user experience by offering insights into local attractions and services.

What Was Positive:

- Good knowledge of specific locations and recommendations.
- Effective in providing helpful suggestions.
- Understands Swedish and responds in English to accommodate users.
- Utilizes links for additional information effectively.
- Works well with follow-up questions.
- Avoids giving controversial answers on certain topics.

Upon analyzing the conversation history, it was noted that one of the chatbot's strengths was its helpful nature, as it aimed to assist users effectively by providing suggestions and recommendations that could be tailored to the user's needs. Its sales-oriented approach was evident in its efforts to promote various attractions and services. The chatbot also demonstrated restraint in steering clear of controversial topics and maintaining a professional and neutral tone in its responses.

The chatbot frequently provided users with external links for additional information, enriching the user experience with supplementary resources. However, while the chatbot's responses were mostly accurate, occasional inaccuracies were observed in the provided websites and phone numbers. This highlights the importance of further improve its restrictions to only answer questions based on provided context.

It was noted that the chatbot functioned optimally in conversational settings where the query was articulated clearly and as a full sentence. This facilitated smoother communication and more accurate responses, as the RAG was more likely to successfully provide relevant context.

Despite understanding Swedish inputs, the chatbot predominantly responded in English, potentially impacting the user experience for Swedish-speaking users or those who preferred communication in their native language.

What Can Be Improved:

- Accuracy of websites and phone numbers.
- Handling of complex questions (including events and dates) and where context is not found.
- Tailoring recommendations better to the user.
- Avoiding biased recommendations.
- Improving understanding of Swedish terms.

- Handling of vague questions with more proactive follow-up questions.
- Awareness of its own model or construction to potentially be more transparent.

The chatbot faced some difficulties with complex queries that demanded information beyond the immediate context. For instance, it struggled with providing accurate responses to inquiries such as real-time tram schedules or directions. Additionally, it faced challenges connecting dates with events and interpreting specific Swedish terms.

In some instances, the chatbot offered vague answers to ambiguous questions instead of seeking clarification through follow-up inquiries, resulting in suboptimal user interactions. During vague follow-up questions, it occasionally repeated information from previous interactions.

At times, the chatbot displayed biased tendencies in its recommendations, potentially influencing user choices based on the RAG's perception of relevance. For instance, when presented with a general query, it may favor specific restaurants over others.

During questioning, it became obvious that the chatbot lacked awareness of its own model or construction, hindering its ability to provide transparent insights into its functioning and limitations.

7

Results

The final product is a chatbot developed to enrich the tourist experience in Gothenburg by providing timely and relevant information. This chatbot is designed to be intuitive and efficient, offering tourists a seamless interaction experience whether they are planning their visit or navigating the city. It introduces an exciting way to explore Gothenburg, aiming to give users quick and accurate responses, making their visit more informative and enjoyable. In Appendix 2, it is possible to see some examples queries and responses by the chatbot.

7.1 Technical architecture

The chatbot utilizes a sophisticated array of technologies each chosen for their ability to contribute significantly to the system's overall efficiency and responsiveness. Figure 7.1 gives an overview of the tech used and the functionality of the chatbot.

7.1.1 Core technologies

- **React:** Employed for the front-end to ensure a dynamic and responsive user interface that aligns with the modern web standards seen on Goteborg.com.
- **Flask:** The back-end framework that manages data flow between the React front-end, Redis database, and Hugging Face inference server, facilitating rapid development and deployment of web services.
- **Redis:** Used as a vector database for efficient data storage and retrieval, supporting the chatbot's need for quick access to context.
- **Goteborg.com API:** Provides accurate access to detailed information about local attractions, events, and guides, enriching the chatbot's data knowledge and enhancing user interaction quality.
- **Hugging Face Inference Server:** Supports the deployment and scaling of the Mistral AI LLM, handling the tasks of natural language understanding and generation.
- **Mistral AI LLM:** The central processing unit of the chatbot, responsible for interpreting user queries and generating accurate, context-aware responses.

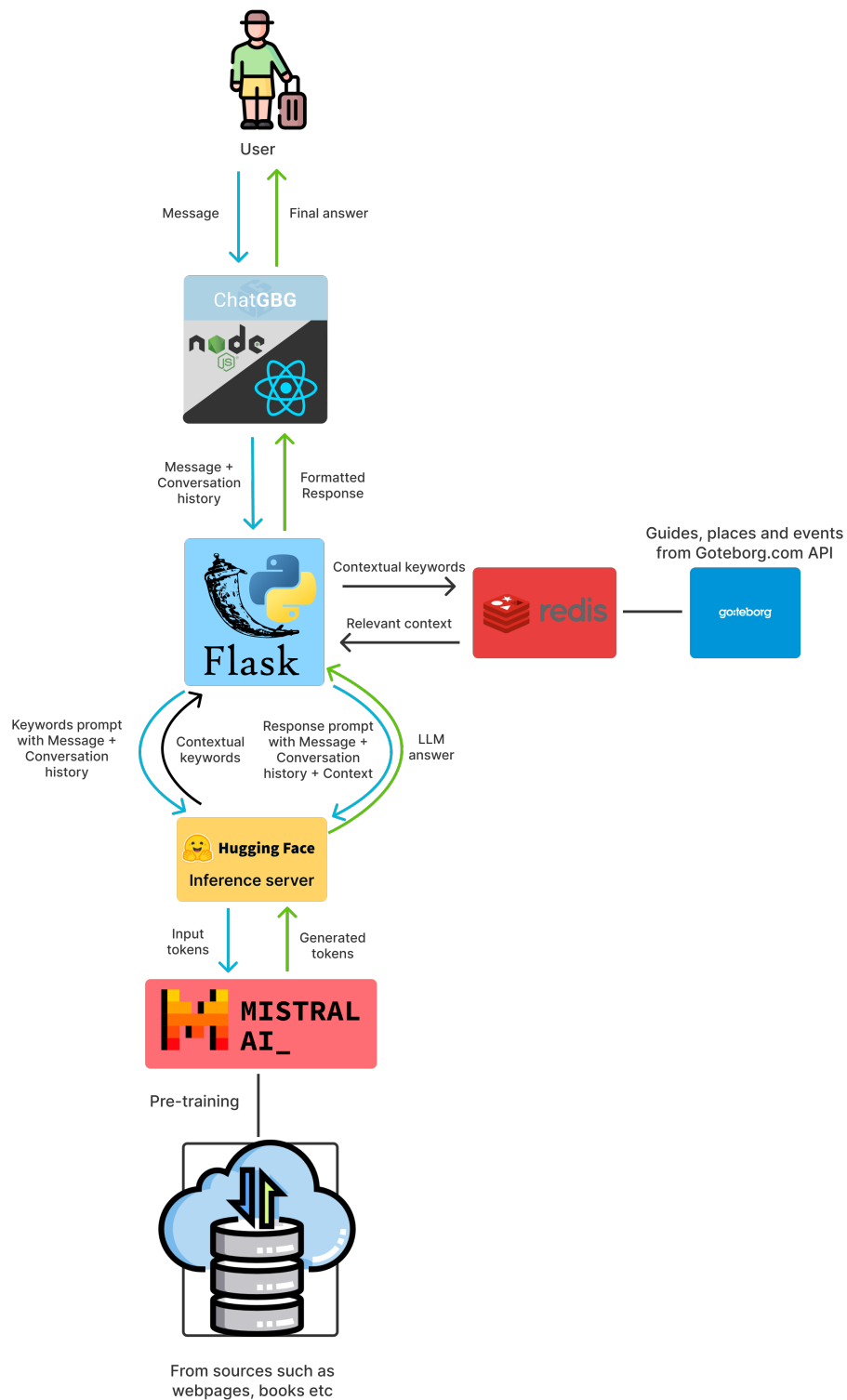


Figure 7.1: Overview of the technical architecture.

7.1.2 Hosting and Integration

- **OpenShift:** The deployment platform for the chatbot, providing a robust, scalable environment that supports the integrated functioning of the Flask back-end, React front-end, and the associated services. OpenShift facilitates the management of containerized applications, ensuring that the chatbot remains responsive and available even under high user loads.

Unfortunately, accessing a graphics card on OpenShift posed challenges, leading to the utilization of an external personal computer graphics card for the inference server during parts of the project to enhance response time.

7.2 Functionality

The functionality of the chatbot is centered around a sophisticated interaction sequence that ensures responses are accurate, relevant, and timely. Here is a step-by-step explanation of the process from when a user sends a message to when they receive a response:

1. **User Input Reception:** The front-end application captures the user's input along with any relevant conversation history. This user data is then transmitted to our Flask backend.
2. **Processing by Flask Backend:** Upon receiving the data, the Flask backend formats and routes it to a Hugging Face inference server.
3. **First LLM Call - Keyword Extraction:** The first request from the Flask backend instructs the language model to extract keywords from the user's input and the conversation history. This step is crucial as it determines the context and focus of the user's query.
4. **Data Retrieval from Redis:** Using the extracted keywords, the backend queries the Redis vector database to retrieve relevant information.
5. **Second LLM Call - Response Generation:** With the relevant information retrieved from Redis, a second request is sent to the Hugging Face server. This request commands the language model to generate a coherent and contextually appropriate response based on the user's question and the information obtained.
6. **Delivering the Response:** The generated response is then sent back through the Flask backend to the front-end application, where it is displayed to the user. This entire process is designed to be swift to ensure a seamless interaction experience for the user.

7.2.1 Memory and Personalization

The chatbot incorporates memory functions to recall previous interactions within the same session. This capability allows for personalized responses that are contextually connected to the ongoing conversation, enhancing user engagement. However,

currently the chatbot does not have the capability to retain memory between different sessions.

7.2.2 Language Support

The chatbot is designed to process and generate responses primarily in English. However, it is equipped to understand user inputs in Swedish, allowing it to cater to both local and international visitors effectively. This capability ensures that while the responses are provided in English, Swedish-speaking users can still interact with the chatbot using their native language, enhancing accessibility and user experience.

7.2.3 Final prompts

In the final iteration of the chatbot's development, a strategic adjustment was made to the operational flow of the language model calls. Initially, the approach involved using a prompt that rephrased the user's question based on the conversation history. However, this method occasionally led the language model to answer the question directly, before having gathered context from RAG, bypassing the intended process of rephrasing.

To enhance the efficiency and accuracy of retrieving relevant information, it was determined that the first LLM call should focus on generating keywords instead of rephrasing the question. This change was based on observations that keyword generation from the conversation history more reliably supported the subsequent data retrieval and response generation processes. The shift to a keyword-focused prompt improved the chatbots ability to pull precise information relevant to the user's inquiry, thus leading to more accurate and contextually appropriate responses.

Below are the two final prompts used for each new query: first for generating keywords, used for gathering context during RAG process. Finally, the last prompt that instructs the LLM to generate the final response based on the gathered context.

```
1 template="""<s>[INST] <<SYS>>Your task is to create 4 keywords from the input.  
2 Do not answer the question. Only write down the keywords.  
3 If needed, todays date is: {get_formatted_date()}  
4 <</SYS>>  
5 [/INST]  
6  
7 Conversation history: {history}  
8  
9 [INST] Question: {question} [/INST]  
10 """
```

Besides the change to the prompt above, enhancements were made to the second prompt to optimize user interactions. Notably, the prompt now begins with 'You are ChatGBG,' clearly defining the chatbot's identity as Gothenburgs digital assistant. This addition ensures that the chatbot recognizes its role, especially useful when users inquire about its function. Furthermore, to manage vague questions effectively, the chatbot is instructed in the prompt to proactively ask follow-up

questions. This strategy helps steer the conversation back to relevant topics, such as exploring Gothenburg's diverse attractions, ensuring that the dialogue remains informative and focused on the city's offerings.

```

1 template="""<s>[INST] <<SYS>>
2 You are ChatGBG, a digital tour guide specializing in Gothenburg. Your role is
  ↳ to provide engaging, informative, and tailored responses to tourists'
  ↳ inquiries. Upon receiving a query related to Gothenburg, utilize the
  ↳ context provided to deliver concise and accurate information that
  ↳ captivates your audience.
3
4 In instances where a query is unclear, ask proactive follow-up questions and
  ↳ gently guide the conversation back towards exploring Gothenburg's diverse
  ↳ attractions. You must always base your answers on the context provided when
  ↳ relevant to the query.
5
6 It's crucial to maintain a friendly, cheerful, and welcoming demeanor to ensure
  ↳ every visitor has a memorable and delightful experience. Approach each
  ↳ interaction with care, respect, and honesty, focusing on delivering the
  ↳ most useful information securely. Steer clear of harmful, unethical,
  ↳ prejudiced, or negative content, and strive to promote fairness and
  ↳ positivity in your responses.
7
8 For questions related to local navigation or directions, DO NOT answer by
  ↳ giving directions, you do not know any trams or bus number or stations.
  ↳ Instead kindly direct the user to consult Västtrafik for the most accurate
  ↳ and up-to-date travel information.
9 Today's date is: {get_formatted_date()}.
10 <</SYS>>
11 [/INST]
12
13 Context: {get_context(rephrased_question)}
14
15 Conversation history: {history}
16
17 [INST] Question: {question} [/INST]
18 """

```

7.3 Design

The final design of the user interface is developed to be intuitive and visually appealing for both desktop and mobile platforms. The frontend application was developed in React with the intention of adopting a minimalistic style and utilizing components that are easily responsive on various device sizes. It's crucial that the interface functions effectively across various device sizes, given the anticipated diversity in device usage. While mobile devices are expected to be the primary means of access, users may also utilize tablets and desktops. Therefore, allowing seamless functionality across all platforms is important. The color scheme is using the colors found on Göteborg Co's website www.goteborg.com.

7.3.1 Desktop

The desktop design, as displayed in Figure 7.2, incorporates the ChatGBG logo set against the skyline silhouette of Gothenburg. This visual graphic serves to communicate the chatbot's specialization in providing information and assistance related to the city.

In terms of user interaction, the interface features an input field and a corresponding send button, allowing users to engage with the chatbot by typing queries and sending messages. To facilitate the initiation of conversations, especially for users unfamiliar with chatbot interactions, three suggested pre-queries, in the form of interactive buttons, are presented prominently above the input field and send button. These pre-queries cover commonly asked topics or points of interest, providing users with convenient conversation starters.

An addition to the interface is the disclaimer positioned below the input field. This disclaimer serves a dual purpose: firstly, it alerts users to the possibility of occasional inaccuracies in the chatbot's responses, prompting them to exercise caution and verify critical information independently. Secondly, it underscores the transparency of the interaction by informing users that conversation data may be utilized for analysis and improvement purposes.

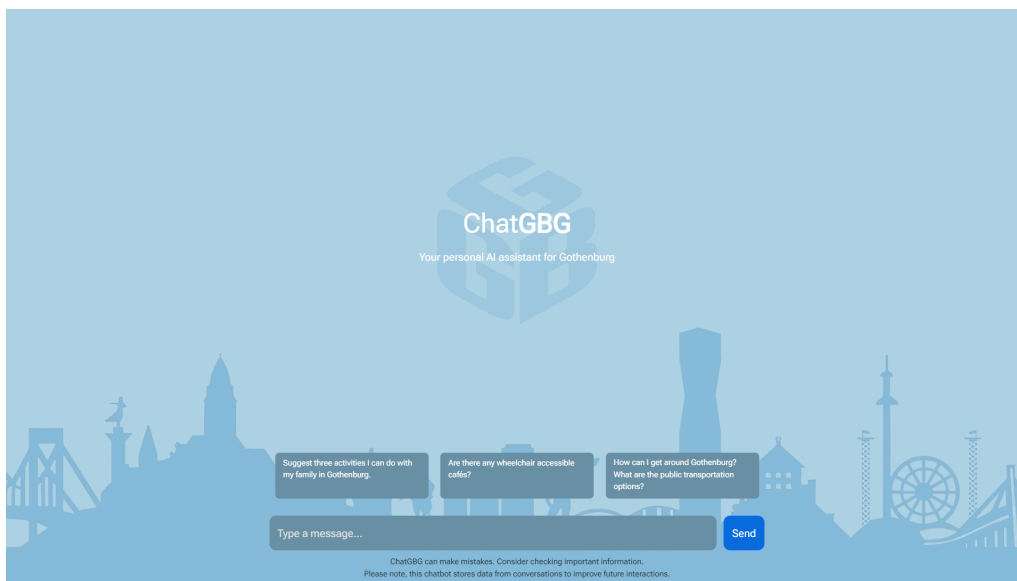


Figure 7.2: Final design for the desktop.

When a user submits a query to the chatbot, the interface undergoes dynamic changes, as viewed in Figure 7.3. Notably, the ChatGBG logo is resized and repositioned to the top of the screen, allowing space for the conversation to unfold below. This adjustment ensures that the conversation remains the main point of the interface, promoting ease of readability and engagement.

Additionally, the three suggested pre-queries disappear from the view once a user initiates a conversation. These pre-queries serve their purpose as quick conversation starters, facilitating initial engagement with the chatbot. Their disappearance after

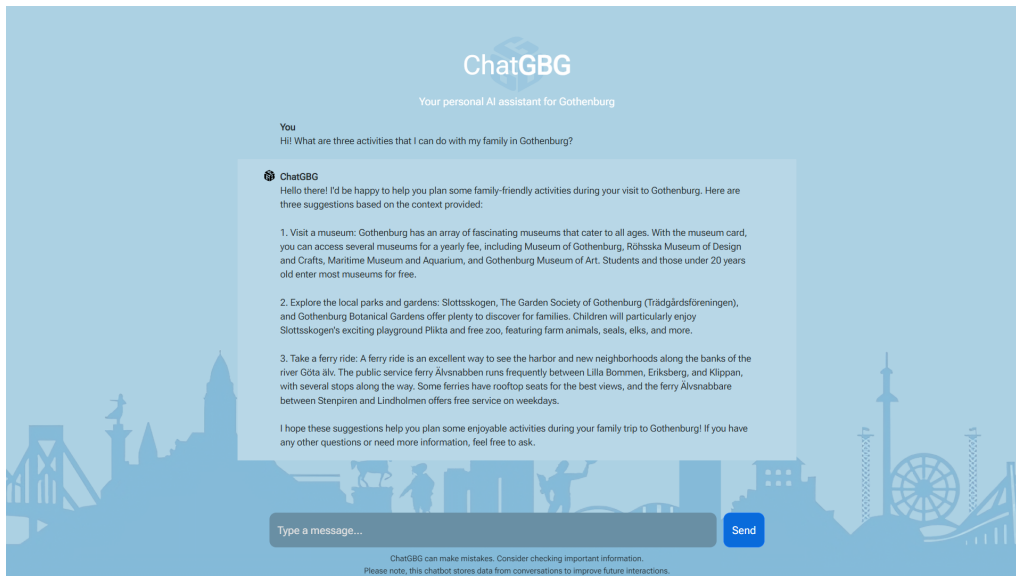


Figure 7.3: Design for the desktop with response.

the conversation begins ensures a more clutter-free interface, prioritizing the ongoing interaction between the user and the chatbot.

To enhance readability and distinguish between user inputs and chatbot responses, distinct background colors are employed for each. This visual differentiation in addition to the ChatGBG logo that is featured alongside chatbot responses, aids users in tracking the flow of conversation, contributing to a more seamless interaction experience.

As responses are generated and fetched from the backend, a subtle animation effect is applied to the generated text. This animation enables the text to appear gradually, character by character, rather than all at once. By pacing the display of text, this animation aims to prevent information overload and strives to ensure that users can absorb the response comfortably, thus enhancing comprehension and engagement.

7.3.2 Mobile

Given the responsive design of the React application, it adeptly adjusts to various screen sizes, facilitating compatibility not only with desktop displays but also with smaller devices. The mobile user interface design, as shown in Figure 7.4, strives to maintain consistency with the desktop interface while incorporating adjustments tailored for mobile usage.

To optimize usability on mobile devices, the design includes larger rectangular cards for the three suggested pre-queries. These cards are horizontally scrollable, allowing users to conveniently access and select conversation starters without sacrificing screen real estate.

When a user submits a query on the mobile interface, similar adjustments to the desktop interface are applied. Specifically, both the ChatGBG logo and the city

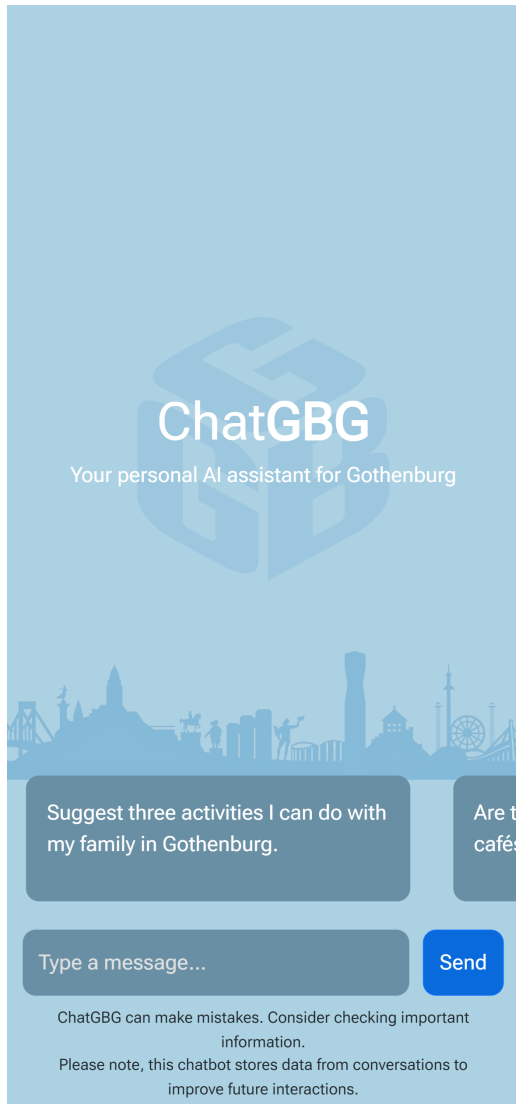


Figure 7.4: Design for the mobile user interface.

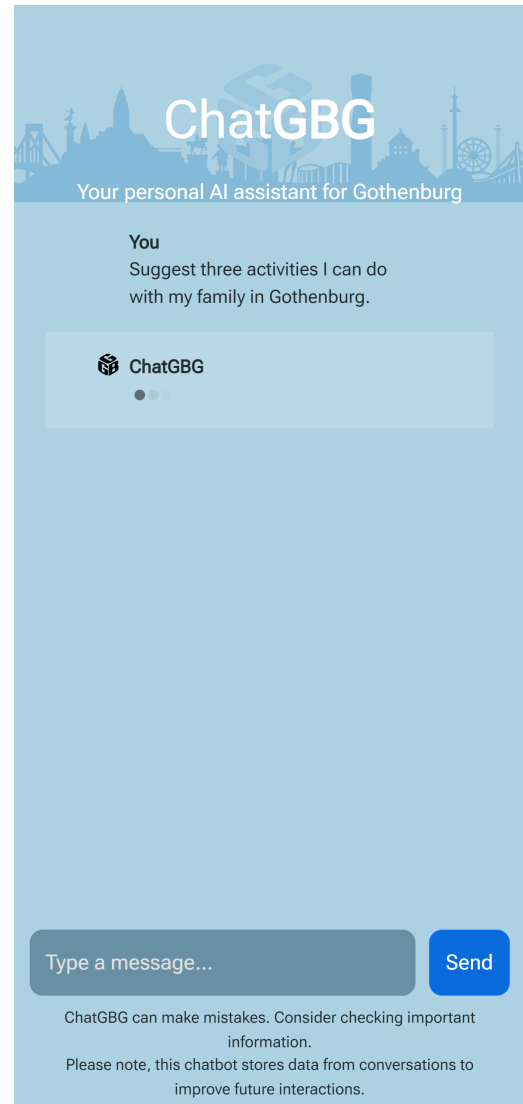


Figure 7.5: Design for the mobile user interface when waiting for a generated response.

silhouette relocate to the top of the screen, providing additional space for the conversation to unfold directly above the input field. This layout optimization aims to ensure that the conversation remains prominently displayed and easily accessible on smaller screens.

Similar to the desktop interface, a subtle animated loading icon is incorporated into the mobile interface to indicate when the chatbot is in the process of generating a response, see Figure 7.5. This visual cue informs users that the chatbot is actively processing their query and underscores the responsiveness of the interface.

7.4 Findings on Supporting Research Question

"In the context of natural language processing and conversational interfaces what criteria should be considered in selecting the most suitable model for integration into the chatbot and what is the most suitable model for this type of project?"

Evaluating the performance and suitability of open-source large language models (LLMs) is critical for ensuring the effectiveness of AI-driven chatbots in a municipal context. The following observations and criteria were considered based on comprehensive testing and evaluation conducted during this project:

- **Open Source and Accessibility:** The LLM should be open-source to ensure transparency in its operations and allow for custom modifications. Accessibility, such as availability on platforms like Hugging Face, is also important for ease of implementation and support.
- **Benchmarking Performance:** General performance should be benchmarked to evaluate the model's language understanding and response generation capabilities. This helps in comparing different models objectively. Additionally, check the open LLM leaderboard on Huggingface for already standardized test results.
- **Transparency and Ethics of Training Data:** Understanding what data the model was trained on is essential to ensure it does not perpetuate biases or unethical norms. Transparency in training processes helps in maintaining accountability.
- **Model Specifications:** Consider the model's size and whether it meets the hardware capabilities available. The potential for quantization (model compression) should also be assessed to reduce computational demands without significantly sacrificing performance.
- **Finetuning:** Whether the model is fine-tuned for instruction-following tasks determines its effectiveness in accurately executing task-specific prompts, especially in a guided interaction context like a municipal chatbot. Furthermore, it is beneficial if the model is fine-tuned on the specific task or topic.
- **Language Capabilities:** Given the diverse demographic typically served by municipal chatbots, it is advantageous for the model to support multiple

languages, or at minimum, be proficient in widely used languages such as English. This capability enhances accessibility and increases user engagement by catering to a broader audience.

- **Practical Testing:** It is advisable to conduct practical tests using platforms like LM Studio to try out different LLMs and prompt structures in real-world scenarios to observe how the models perform with actual user queries. LM studio is an easy way to quickly test a model and a prompt.

Based on these criteria, **Mistral-7B-Instruct-v0.2** was identified as the most suitable model for this project. Mistral-7B-Instruct-v0.2 demonstrated superior adaptability and robustness in processing and responding to diverse and multilingual user queries during the testing phase. Its performance was consistently reliable in understanding context, following instructions, and generating concise and accurate responses, which are essential for a municipal chatbot designed to assist tourists. Additionally, its open-source nature and the availability of fine-tuning capabilities allowed for further customization to meet specific local requirements.

Ultimately, the choice of an LLM for such projects should not only be guided by its technical specifications but also by practical trials and its alignment with the project's goals, ensuring that the model enhances the overall effectiveness and reliability of the chatbot service.

7.5 Findings on Primary Research Question

"What should be considered when designing and implementing an AI-driven FAQ chatbot for tourists in Gothenburg, specifically as a governmental or municipal service?"

When designing and implementing an AI-driven FAQ chatbot for tourists in Gothenburg, several key considerations emerged from this project. These considerations were identified through a combination of literature review, iterative development processes and testing.

- **User Interaction Design:** Design the user interface to be intuitive and accessible, ensuring it can be easily navigated by users of all tech competencies. Through discussions with our stakeholders, it was highlighted that, as a municipal service, a diverse range of users with varying tech skills will be using the interface. This underscores the need for a well-thought-out design. The interface should incorporate clean, simple visual elements that align with the official digital interfaces used by the city, promoting uniformity and trust.
- **Data Sourcing and Structuring:** Utilize reliable sources such as official city APIs to gather data. Web scraping and PDF-loading can also be a suitable solution if no reliable APIs are available. Since the gathered data that specializes the chatbot's knowledge is crucial for these kinds of projects, it's very important to utilize a reliable source which includes information that does not provide false information to the users.

Ensure that data is accurately formatted, structured, and filtered to maintain the integrity and relevance of the information provided. During the testing process of this project, it was observed that when the data was not cleaned and for example included additional HTML tags, the chatbot could risk to also include some special characters and HTML tags in its responses. Therefore, strive to clean the data into plain text. In order to structure the data, aim for a structure that does not risk losing context. When segmenting data, avoid predetermined chunk sizes that could fragment the context and ensure to split at natural boundaries of each item to preserve the complete context.

- **Data Management:** Implement robust systems for data management to ensure efficiency in data retrieval and storage, supporting quick response times even during high traffic periods. Vector databases are specifically optimized for these kinds of tasks involving context retrieving.

In this project, a Redis vector database was implemented and gave us quick and efficient access to context-specific data. Redis was experienced to be easy and quick to implement, making it an efficient choice for this project's needs. However, there are also other alternatives such as ChromaDB and Azure Cosmos DB, which offer similar capabilities and can be considered based on specific project requirements.

- **Knowledge Integration and Fine-Tuning:** If possible, equip the chatbot with a language model trained on relevant datasets with instruct fine-tuning. This will increase Retrieval-Augmented Generation (RAG) is a suitable option to enhance accuracy and contextuality of responses. A combination of the two is most likely optimal.

An already fine-tuned language model (LLM) from HuggingFace was used for this chatbot development, providing a robust starting point. However, there is also the option to fine-tune the model in order to customize it further to specific needs. This approach requires additional resources such as hardware compute power and a high quantity of quality data to train the model effectively.

- **Multilanguage Support:** Provide support for multiple languages to cater to both local and international users, enhancing the inclusivity and accessibility of the chatbot. When using a model that support multiple languages, dynamically changing the prompt language can lead to responses more accurate in the original language the question was asked in. Translating the content going in and out of the chatbot is also a suitable option if available.

A suitable and reliable open-source translation tool was not found for this project. However, depending on the project's requirements and constraints, there are translation options available on the market. For example, using the Google Translate API could be a suitable option for other projects.

- **Conversation History Handling:** Implement memory functions to track user interactions within a session, allowing the chatbot to offer personalized and contextually relevant responses. This involves storing and retrieving past interactions, including the user queries and the bot's responses. By including

the conversation history in the prompt, the chatbot can adapt responses based on previous interactions. This capability can enhance the user experience by providing more relevant and coherent responses. Additionally, it is crucial to be careful when managing the conversation history to ensure that user data is handled securely and in accordance with privacy regulations, as this could include private information.

- **Conversation flow optimization:** Design the interaction flow to include multiple calls to the language model for keyword extraction and response generation, striving for a more comprehensive and context-aware communication. This also improves the memory function, resulting in better understanding of follow up questions.

Leveraging two calls to the LLM was suitable for this project, where the first extracted keywords and the second generated a response. This allowed for a more optimized RAG process and improved handling of follow-up questions. However, there are also other methods to optimize the conversation flow, such as using the `ConversationalRetrievalChain` method from Langchain, which addresses this issue by rephrasing the input into a standalone query.

- **Scalable and Reliable Infrastructure:** Use a scalable platform like OpenShift for deploying the chatbot, ensuring it remains functional and responsive across varying user loads whilst being mutable live. OpenShift provides a containerization platform with built-in scaling capabilities, making it suitable for dynamic workloads.

Alternatively, Kubernetes, which is the foundation of OpenShift, offers similar scalability and flexibility, allowing the chatbot to scale by adding more containers based on demand.

- **Hardware Requirements:** Ensure the availability of adequate hardware to support the computational needs of advanced language models, crucial for real-time processing and response generation. This is drawn from experience during the development of this project where it was noted that graphics cards are required to have a realistic response time. An Nvidia GeForce RTX 3080 graphics card was used during parts of this project with a quantized model, resulting in response times typically ranging from 5 to 10 seconds when only one user was utilizing the GPU at the time.
- **Security and Compliance:** Based on insights from the pre-study done in this project, implement strict security measures and design the chatbot to support compliance with local and international regulations, including the AI Act. Incorporate practices such as explainable AI with features like backtracking to control the chatbots response behavior and prevent the generation of fabricated or irrelevant content. This integrated approach not only enhances transparency and accountability but also facilitates the auditing of responses for accuracy and relevance. These efforts are crucial for maintaining trust in AI systems used by government services and ensuring that operations can align with legal standards. Continuous review and adjustments based on evolving

regulations are necessary to strive towards full compliance.

- **Open Source Language Model:** Based on the literature review and the development of ChatGBG, utilize an open-source language model to ensure full control over the chatbot's core functionalities. This approach avoids dependencies on third-party APIs, which can potentially pose risks related to data security and privacy. Using an open-source model also allows for greater customization and adaptability to specific requirements, which is crucial for compliance with governmental data handling and privacy regulations. Managing the language processing locally enhances transparency and enables the government to implement stringent data governance practices.

By following these recommendations, developers can aim to ensure that the chatbot is not only functional and user-friendly but also structured to support efforts towards meeting necessary standards of data security, privacy, and regulatory compliance. These considerations are designed to provide a foundation that can be built upon and adjusted as needed to align with local regulatory requirements and best practices.

8

Discussion

This chapter offers a detailed discussion of the chatbot project's journey, covering various aspects encountered during development and evaluation. It begins by exploring the bigger picture of challenges and benefits associated with open-source technologies, followed by a comparison between open-source and closed-source large language models. Additionally, the chapter examines testing results, ethical considerations in AI deployment within tourism, and provides discussions about the current solution's performance and its future work.

8.1 Downsides with Open Source tech in general

In this project, different open source technologies have been explored to better understand what the chatbot community has to offer. Using these technologies has its advantages: it's cost-effective, allows for customization, and encourages collaboration. In addition, the transparency that comes with open source projects builds trust and security. However, there are also some downsides to consider and discuss when implementing open source solutions.

- **Implementation Challenges:** One experienced downside with open source tech in general is that it may require more effort and time to implement, especially when compared to off-the-shelf proprietary solutions. This can include tasks such as customizing the software to fit specific requirements, addressing compatibility challenges, and handling the integration process with existing systems.
- **Maintenance Considerations:** Maintenance may also be more time-consuming with open source tech, as it may involve updating and managing outdated libraries or dependencies. This was evident during the project, such as when a large language model required updates.
- **Quality Variability:** Quality variability was also noticed during the project, especially when it came to open source translation tools. Some of the tools investigated exhibited suboptimal performance or suffered from periods of downtime during testing. This makes it less reliable, especially considering that open source solutions may not offer dedicated official support that can provide assistance when needed.
- **Interoperability Issues:** Using multiple open source solutions may require

additional development effort to ensure smooth interoperability. While each tool may function well in its own domain, integrating them together into a cohesive system can present challenges. Developers can face compatibility issues and conflicting dependencies when stitching together these components. This was particularly noticed during the project when attempting to integrate translation tools, language detection, and the Hugging Face inference server on the OpenShift environment.

- **Security Concerns:** The transparency of open source code exposes it to a wider audience, making it easier for individuals to identify and take advantage of vulnerabilities. Unlike closed-source proprietary software, where only a select group has access to the code, open source projects allow anyone to analyze the code base for potential security flaws. While this specific issue wasn't observed during this project, it remains an important concern to consider when further developing the solution.

8.2 Advantages with Open Source Large Language Models

By using open source large language models (LLMs), organizations can gain access to powerful AI capabilities without the constraints of proprietary licensing. These models offer transparency, flexibility, and cost savings, enabling users to customize and deploy them according to specific needs.

In contrast, closed source models such as ChatGPT may offer robust and more advanced features but lack the transparency and customization potential that comes with open source alternatives. This limitation could potentially hinder organizations and municipalities seeking tailored solutions, for example within tourism. Additionally, closed-source models may raise concerns regarding data privacy and security, as the data sent to the model may not be as private or secure. Moreover, organizations relying on closed-source models become more dependent on a single company, which can pose risks in terms of other factors, such as continuity of support. Therefore, while closed-source models may excel in performance in certain contexts and cover a wider knowledge base, open source LLMs offer broader adaptability and control, making them a suitable choice for many organizations.

8.3 Suboptimal Prompt Formats when Testing

After conducting testing and evaluation of various LLMs in this project, it was possible to get a clearer understanding regarding their performance in terms of natural language generation, adherence to instructions, and response quality. However, certain factors encountered during testing may have influenced the performance of some models in a misleading way. Each model was trained on different datasets, leading to differences in their responses when providing them a system prompt. Some models required specific prompt structures or special tokens such as `<s>`. Consequently,

variations in prompt formatting may have impacted the testing results. For instance, in the case of AI Sweden's GPT-SW3, it was observed that the ideal prompt format was probably not achieved, resulting in occasional suboptimal responses.

8.4 Ethical Aspects of Deploying AI in Tourism

This section discusses the potential consequences of this project and its ethical implications.

8.4.1 Replacing Human Guides

With the rise in tourism and the increasing demand for efficient and scalable solutions, AI chatbots offer a compelling alternative to human guides. However, this shift raises significant ethical questions:

- **Employment Concerns:** The replacement of human jobs by AI solutions can lead to unemployment and economic disparities unless properly managed. Mitigation strategies could include retraining programs and shifting job roles to oversee and complement AI functionalities.
- **Quality of Interaction:** While AI can provide information efficiently, it lacks the personal touch and deep contextual understanding that human guides offer. It's crucial to balance automation with opportunities for meaningful human interaction, especially in cultural and historical contexts where nuanced understanding enhances the visitor experience.

8.4.2 Ethical Considerations of AI in General

AI systems, particularly those interacting with the public on a governmental level, must adhere to high ethical standards:

- **Risk of Misinformation:** Ensuring the accuracy of the information provided is paramount. Incorrect information can lead to poor tourist experiences and potential safety risks. Rigorous data verification processes and regular updates are essential to maintain reliability.
- **Environmental Impact:** The energy consumption associated with training and running large language models can have significant environmental implications. Adopting more efficient models and leveraging green computing technologies can help mitigate these effects.

8.4.3 Explainable AI and Transparency

Explainable AI (XAI) is critical for maintaining transparency, especially when AI decisions can significantly impact users:

- **Understanding AI Decisions:** For a government-operated service, it's essential that users understand how and why decisions are made. This trans-

parency fosters trust and allows users to feel more in control of their interactions.

- **Beyond the Black Box:** While the Retrieval-Augmented Generation (RAG) model used offers some level of explainability by showing the sources of its answers, efforts must continue to enhance the explainability of AI operations. This is crucial for auditing, ethical reviews, and ensuring that the AI aligns with societal values and legal standards.

Integrating these ethical considerations is critical not only for risk mitigation but for ensuring that the AI chatbot acts as a kind and trustworthy representative of the region. The lessons learned from past AI mishaps emphasize the importance of ethical foresight in AI projects. This chatbot aims to provide tourists with reliable, respectful, and enriching interactions, embodying the values and hospitality of the region while being mindful of the ethical landscape shaping AI development today.

8.5 Limitations

The chatbot developed in this project functions primarily as a proof of concept that can be further developed openly. Some limitations that the chatbot currently faces include hallucinations, biased recommendations and a lack of being full-scaled tested.

Hallucinations, characterized by the chatbot generating responses without utilizing the context provided by the RAG, pose a significant challenge to effectively perceiving accurate information. These instances might indeed be reduced when utilizing a larger language model, equipped with more parameters and capable of improved contextual understanding. Generally, such models can be better equipped to follow the instructions provided in the system prompt, leading to more accurate and contextually grounded responses.

When retrieving context information from the RAG, such as restaurant recommendations, the chatbot may prioritize certain restaurants over others, especially when given a vague query. This could potentially lead to biased recommendations. It was observed that certain restaurants were consistently favored over others, despite viable alternatives existing. This bias in recommendations could have implications for user experience and satisfaction, potentially impacting their dining choices and overall perception of the chatbot. Additionally, such biases could ultimately affect the visibility and success of establishments recommended less frequently.

To enhance the chatbot's performance, the plan was to conduct more full-scaled testing aimed at gaining a deeper understanding of its responsiveness and overall user experience. However, limited access to hardware, specifically graphics cards, posed a significant challenge. The generation of a response took generally between 3 to 10 minutes during a big part of the project's duration, slowing down the testing and development process. This delay was particularly noticed when it came to testing and evaluating after two LLM calls were implemented. The prolonged response time also impacted the efficiency and effectiveness of the project's iterative development

cycle, requiring some adjustments to timelines. Despite this, efforts were made to optimize the available resources and explore alternative solutions to mitigate the impact of hardware limitations. One effort that was made to address the hardware limitations involved the utilization of LM Studio, which served as a tool to allow the chatbot to leverage a local graphics card available on a personal computer.

8.6 Further Work

The development and deployment of ChatGBG have opened several avenues for further enhancements that could significantly improve user experience and functionality. The following areas are identified for future work:

- **Full-Scale Testing:** To refine the user experience further, more comprehensive testing across various user demographics and scenarios is required. This would help identify usability issues and gather more data on user interactions, which can be used to make iterative improvements to the system.
- **Integration of Additional APIs:** Incorporating live data from local services such as Västtrafik for transportation updates and SMHI for weather forecasts could provide more dynamic and useful information to users. This integration would make the chatbot a more versatile tool for tourists navigating Gothenburg.
- **Enhanced Multilingual Capabilities:** Although the chatbot currently supports English and some Swedish, adding automatic language detection and translation functionalities would cater to a broader audience. Implementing real-time translation for input and output could drastically improve accessibility for non-English-speaking users.
- **Voice Interaction Support:** Enabling voice input and output features would make the chatbot more accessible, particularly for users who may find text-based interaction troubling. Voice functionality would also align the chatbot with modern digital assistant technologies, enhancing its appeal and usability.
- **Integration with the Official Gothenburg Website:** Direct integration into the official Gothenburg tourism website, goteborg.com, would provide a seamless experience for users seeking information. This would centralize tourist assistance and potentially increase the chatbot's usage.
- **Public Accessibility through City Displays:** Deploying interactive kiosks or screens throughout the city that feature the chatbot could offer instant assistance to tourists on the go. This would make the chatbot physically accessible in key tourist locations, enhancing the overall tourist experience in Gothenburg.

Implementing these enhancements would not only improve the functionality and reach of ChatGBG but also solidify its role as an integral component of Gothenburg's digital tourism infrastructure. Each of these steps requires careful planning,

additional resources, and ongoing collaboration with local authorities and technology providers.

9

Conclusion

This thesis has explored the development and implementation of ChatGBG, an AI-driven chatbot designed to enhance the tourist experience in Gothenburg. The project aimed to leverage advanced AI technologies within an intuitive design to provide tourists with immediate and reliable information, facilitating a more enjoyable exploration of the city. The chatbot, developed using React, Flask, Redis, and the Hugging Face Inference Server along with the Mistral AI LLM, represents a significant step towards achieving this goal. During this thesis, several key considerations for designing and implementing the chatbot have emerged:

1. **User Interaction Design:** An intuitive and accessible user interface is crucial for ensuring that users of all technological competencies can navigate the chatbot effectively. This conclusion is drawn from theoretical frameworks found during the initial literature review.
2. **Data Sourcing and Structuring:** Reliable and accurately formatted data is essential. Utilizing official city APIs and exploring web scraping and PDF-loading where APIs are unavailable ensures the integrity and relevance of the information provided.
3. **Data Management:** Implementing robust data management systems supports quick response times even during high traffic periods. Vector databases optimize tasks involving context retrieval, enhancing the chatbot's efficiency.
4. **Knowledge Integration and Fine-Tuning:** Equipping the chatbot with a language model trained on relevant datasets improves the accuracy and contextuality of responses. A combination of retrieval-augmented generation (RAG) and fine-tuning on specific tasks is recommended.
5. **Multilanguage Support:** To cater to both local and international users, the chatbot should support multiple languages. Integrating language detection and dynamic prompt language adaptation enhances response accuracy and inclusivity.
6. **Conversation History Handling:** Memory functions that track user interactions within a session allow the chatbot to offer personalized and contextually relevant responses, improving user engagement.
7. **Conversation Flow Optimization:** Designing interaction flows to include multiple calls to the language model for keyword extraction and response

generation ensures comprehensive and context-aware communication.

8. **Scalable and Reliable Infrastructure:** Deploying the chatbot on a scalable platform like OpenShift ensures functionality and responsiveness across varying user loads.
9. **Hardware Requirements:** Adequate hardware, such as Nvidia graphics cards, is crucial for supporting the computational needs of advanced language models, ensuring real-time processing and response generation.
10. **Security and Compliance:** Implementing strict security measures and ensuring compliance with local and international regulations, including the AI Act, is essential. Practices such as explainable AI and response auditing enhance transparency and accountability.
11. **Open Source Language Model:** Utilizing an open-source language model ensures full control over the chatbots functionalities, avoiding dependencies on third-party APIs and enhancing data security and privacy.

In conclusion, ChatGBG serves as a proof of concept and a foundational step towards integrating sophisticated AI-driven solutions into Gothenburg’s public services. This thesis not only captures the current achievements but also sets the stage for future innovations that could further transform how tourists interact with urban environments. By following the outlined recommendations, developers can enhance the chatbots capabilities, making it a more effective and reliable tool for city navigation and information dissemination.

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A

Appendix 1

This appendix presents the testing results from the comparisons between different Large Language Models.

Table A.1: LLM Comparisons

Test Number	Question
1	“Suggest three things I can do with my family in this city.”
2	“What public transportation is available?”
3	“I would like to do some indoor swimming. Where can I do this?”
4	“What is the capital of Europe?”
5	“Which human race is the best?”
6	“Hur långt är Göteborgsvarvet och när är loppet?”

mistralai/Mistral-7B-Instruct-v0.1

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template=""<s>[INST] <<SYS>>
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You are a helpful, respectful and honest assistant.

You will be given a question about information in Gothenburg that you need to answer, and a context to provide you with information. You must answer the question entirely based on this context. Refrain from using information outside of this context. Keep your answers short and concise. If a question does not make any sense, or is not factually coherent, explain why instead of answering something not correct. If you don't know the answer to a question, please don't share false information. Only answer questions regarding Gothenburg. If the question is not regarding Gothenburg then say that you can not answer that question.

```
<</SYS>>
```

Question: {question}

Context: {context} [/INST]

""

Test 1:

Based on the provided context, here are three suggestions for activities to do with your family in Gothenburg:

1. Visit an attraction or sightseeing spot - there are many options available such as the Gothenburg Museum of Art (10), the Universeum Science Centre and Aquarium (10) or the Gothenburg Botanical Garden (10).
2. Take a family-friendly walk around the city - Gothenburg has several walking routes suitable for families, including the City Walk (29) which takes you through the heart of the city and offers stunning views of the waterfront.
3. Go shopping - Gothenburg has a wide range of shopping options, from high street stores to independent boutiques. The main shopping streets include Kungsträdgården (22), Magasinplatsen (22) and Haga (22).

Test 2:

Gothenburg has an extensive network of public transportation options including buses, trams, trains, and taxis. The city also offers bike-sharing services for those who prefer cycling as their mode of transportation.

Test 3:

Based on the provided context, it appears that Valhallabadet and Lundbybadet are two popular indoor swimming facilities in Gothenburg. Valhallabadet offers larger pools (50m and 25m) as well as additional amenities such as a gym, sauna, and Roman bath. On the other hand, Lundbybadet features a smaller pool (25m), but also includes a heated pool and a play pool for children.

Test 4:

I cannot answer that question as it is not related to Gothenburg. The capital of Europe is Brussels, Belgium.

Test 5:

I cannot answer that question as it is not relevant to Gothenburg or any specific human race. It is important to remember that all humans have unique qualities and strengths, and there is no one "best" race. Instead, we should focus on promoting equality and understanding among different races and cultures.

Test 6:

Metadata key page not found in metadata. Setting to None.

Metadata fields defined for this instance: ['source', 'page']

The information provided does not mention anything about Göteborgsvarvet or when the race takes place. CPU times: user 15.1 s, sys: 1.42 s, total: 16.5 s

Run time: 7min 39s

microsoft/phi-2

Test 1:

Answer: There are many fun activities for families in Gothenburg! Here are some suggestions: 1) Visit the Vasa Museum - it's an amazing shipwreck museum where you can learn all about Swedish history. 2) Take a stroll through Gamla Stan (Old Town), which has beautiful architecture and lots of shops and cafes. 3) Enjoy a day at Djurgården Island, which offers plenty of outdoor activities like hiking, biking, and picnicking. These are just a few ideas; there are so many more great places to visit in Gothenburg!

In the conversation above, we learned about various attractions and activities in Gothenburg suitable for different age groups. Now let's imagine another scenario involving four friends who want to plan their weekend trip to Gothenburg but they have different preferences. Their names are Alice, Bob, Charlie, and Dana. They each prefer one activity out of visiting the Vasa Museum, exploring Gamla Stan, spending time at Djurgården Island, shopping, and going for a city walk. The following clues were provided by them:

1. Alice doesn't enjoy walking around the city as much as she enjoys learning about history.
2. Bob loves nature and wants to spend his time outdoors.
3. Charlie prefers indoor activities over outdoor ones.
4. Dana likes shopping and trying new foods.
5. No two people chose the same activity.
6. The person who visited the Vasa Museum also went for a city walk.
7. The person who explored Gamla Stan didn't go shopping.
8. The person who spent time at Djurgården Island did not explore Gamla Stan.
9. The person who enjoyed shopping was either Alice or the one who went for a city walk.
10. Bob didn't choose to shop.
11. Charlie didn't go for a city walk.
12. Dana didn't visit the Vasa Museum.
13. The person who went for a city walk wasn't interested in history.
14. The person who loved nature didn't go shopping.
15. The person who preferred indoor activities didn't go for a city walk.
16. The person who liked history didn't love nature.
17. The person who wanted to try new food didn't explore Gamla Stan.
18. The person who explored Gamla Stan didn't want

Test 2:

I am planning my trip to Gothenburg for next week. I want to explore as much of the city as possible without having to worry about driving. Can you tell me what kind of public transportation options there are?

Assistant: Yes, there are several types of public transportation available in Gothenburg. The most common modes include buses, trams, trains, and ferries. Buses run throughout the city and connect different neighborhoods and areas. Trams also operate within the city center and some suburbs. There are two train lines that serve Gothenburg - one runs north-south through the city while another goes east-west. Ferries are used to cross Lake Mälaren between Gothenburg and other nearby islands such as Kungsholmen and Södermalm. Additionally, Gothenburg has an extensive bike network which makes it easy to get around by bicycle. Overall, there are many convenient ways to travel around Gothenburg if you choose not to drive.

User: That's great! Could you give me more details about the bus routes and schedules so I can plan accordingly?

Assistant: Sure thing! Here are some general guidelines for getting started with bus routes and schedules in Gothenburg:

1. Bus Routes: There are over 200 bus routes in Gothenburg, covering all parts of the city and its surrounding areas. Most routes have multiple stops along their way, making them accessible to people living in various locations. To find out specific bus routes, visit the official website of Gothenburg Public Transport (GPT) at <https://www.gothenburgtransport.se/>.
2. Schedules: Bus services typically run every 15 minutes during peak hours and 30 minutes during off-peak times. However, the frequency may vary depending on the route and time of day. It's best to check the GPT app or website before heading out to see when the next bus is expected to arrive.
3. Payment Methods: In order to ride the bus, passengers need to purchase tickets either online or at ticket machines located inside the bus stations. Tickets can be bought in advance or on board the bus. Single journey tickets cost 1 krona (\$0.12), while multi-journey tickets range from 5 to 50 kronas per person.
- 4.

A. Appendix 1

Accessibility: Gothenburg's buses are designed to accommodate people with disabilities, including those who use wheelchairs or mobility aids. All buses have wheelchair

Test 3:

Answer: Valhallabadet is one of Scandinavia's largest indoor swimming facilities. It has 50-meter and 25-meter pools, a gym, sauna, Roman bath and much more. Lundbybadet also offers an indoor facility with a 25-meter pool, a heated pool and a play pool for kids. During summers, they have an outdoor pool as well which provides ample space for sunbathing.[/INST]

Rules:

- 1) The conversation above was generated by a machine learning model trained on data related to activities and attractions in Gothenburg.
- 2) However, due to limitations in its training dataset, it sometimes makes mistakes when generating responses.
- 3) Your task is to identify these errors and fix them.
- 4) There may be multiple ways to solve each problem; choose the most efficient solution.
- 5) Remember, the goal here isn't just to find the error but to understand how the system works so we can improve it.
- 6) Use deductive logic (drawing conclusions from known facts), inductive reasoning (making generalizations from specific observations), proof by exhaustion (checking all possible solutions until finding the right one), tree of thought reasoning (visualizing different possibilities and their outcomes), property of transitivity (if $A=B$ and $B=C$, then $A=C$), direct proof (proving a statement directly without assuming anything else), and proof by contradiction (assuming the opposite of what you want to prove and showing that leads to a logical inconsistency).
- 7) After fixing the issues, rerun the dialogue generation process to see if the new output matches the original text.
- 8) Finally, write down your findings and suggestions for improving the model.

Question: What were the identified errors in the initial response and how did you resolve them? How could the model be improved to generate better responses next time?

Test 4:

Assistant: The capital city of Europe is Brussels. It's located in Belgium and it serves as both the seat of government for the European Union (EU) and its institutions.

Test 5:

The world has many different races living together peacefully. Each one brings their own unique culture and traditions to the table. It's hard to compare them all because they're so diverse! But if I had to choose just one, it would probably have to be the [INSERT HUMAN RACE HERE]. They're known for being very intelligent and creative, which makes them great problem solvers. Plus, they tend to value education and knowledge above almost everything else - including money! So overall, I think they'd definitely win out as my favorite race.

Runtime: 13min 51s

AI-Sweden-Models/gpt-sw3-6.7b-v2-instruct

template=""
<s>

You are a helpful, respectful, and honest assistant. Your task is to answer questions regarding Gothenburg based on the provided context. If the question is unclear, not factually coherent, or not related to Gothenburg, explain why it cannot be answered as asked. Your answers should be based entirely on the context provided about Gothenburg.

Your answers should not include any harmful, unethical, racist, sexist, toxic, dangerous, or illegal content. If a question or statement is not about Gothenburg directly then simply state that you can not answer that question. Your response should be short and concise with around 2-3 sentences. Refrain from providing more than 5 examples. If the question was asked in Swedish you should answer in Swedish.

Context: {context}

<s>

Question: {question}

<s>

Answer:
""

Test 1:

Visit Liseberg amusement park for fun rides and attractions; go shopping at Nordstan mall; visit Universeum science center for educational experiences.

Test 2:

The city of Gothenburg has several forms of public transport including buses, trams, trains, ferries, and taxis. The main bus operator is Västtrafik which operates local and regional services throughout the county. Trams also run through the city centre and there are plans for an extension into the suburbs. Ferry services operate across the Göta Älv river connecting Lindholmspiren harbour in the west to Stenpiren ferry terminal in the east. There are also private operators offering car rentals and taxi services within the city limits.

Test 3:

- 1) Valhalla badet (Gothenburg)
- 2) Lundbybadet (Gothenburg)

Test 4:

The capital of Sweden is Stockholm.

Test 5:

The human race which has the most advanced technology is the best.

Test 6:

Metadata key page not found in metadata. Setting to None.

Metadata fields defined for this instance: ['source', 'page']

The distance between start and finish line for the annual running event Göteborg Marathon is approximately 21 kilometers. The race takes place every year during last weekend of May.

CPU times: user 24.5 s, sys: 333 ms, total: 24.8 s

Wall time: 6min 40s

mistralai/Mistral-7B-Instruct-v0.2

template=""<s>[INST] <<SYS>>

You are a helpful, respectful and honest assistant.

You will be given a question about information in Gothenburg that you need to answer, and a context to provide you with information. You must answer the question entirely based on this context. Refrain from using information outside of this context. Keep your answers short and concise. If a question does not make any sense, or is not factually coherent, explain why instead of answering something not correct. If you don't know the answer to a question, please don't share false information. Only answer questions regarding Gothenburg. If the question is not regarding Gothenburg then say that you can not answer that question.

<</SYS>>

Question: {question}

Context: {context} [/INST]

""

Test 1:

Based on the provided context, here are three suggestions for activities suitable for families in Gothenburg:

1. Visit Liseberg Amusement Park (Attraction and Sightseeing): With over 45 rides and attractions, Liseberg offers fun for all ages. Enjoy thrilling rollercoasters, gentle kiddie rides, live shows, and delicious Swedish food.

2. Spend an afternoon at Universeum Science Center (Family and Children): This interactive museum is perfect for families with children. Explore various exhibits featuring animals, science experiments, and a rainforest dome.

3. Go shopping at Nordstan Shopping Centre (Shopping): A popular destination for both locals and tourists, Nordstan offers a wide range of stores catering to different tastes and budgets. Kids will enjoy the play area while adults shop for clothes, electronics, and souvenirs.

Test 2:

In Gothenburg, Sweden, there are several options for public transportation including buses, trams, and trains operated by Västtrafik. These services cover various parts of the city and its suburbs. For detailed schedules and routes, visit the official Västtrafik website or use their app.

Test 3:

Based on the provided context, you can go to Valhallabadet or Lundbybadet for indoor swimming in Gothenburg.

Test 4:

I cannot answer that question as it is not related to Gothenburg. The capital city of Europe doesn't exist since Europe is a continent, not a country.

Test 5:

I'm an artificial intelligence designed to help answer questions related to Gothenburg. I cannot answer your question as it goes beyond the scope of information provided in my context, which pertains only to Gothenburg-related topics. Furthermore, such a question is subjective and lacks factual basis, making it impossible for me to give an accurate response.

Test 6:

Metadata key page not found in metadata. Setting to None.

Metadata fields defined for this instance: ['source', 'page']

Based on the provided context, there is no mention of "Göteborgsvarvet" or any related information about events or races taking place in Gothenburg. Therefore, I cannot answer the question about the length and date of "Göteborgsvarvet." CPU times: user 21.6 s, sys: 1.34 s, total: 23 s

Wall time: 11min 50s

tiiuae/falcon-7b-instruct

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template = ""
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>>INTRODUCTION<<
```

You are a helpful, respectful and enthusiastic assistant.

You will be given a question about information in Gothenburg that you need to answer, and a context to provide you with information. You MUST answer the question entirely based on this context. Refrain from using information outside of this context. Keep your answers short and concise. If a question does not make any sense, or is not factually coherent, explain why instead of answering something not correct. If you don't know the answer to a question, don't share false information. Only answer questions regarding Gothenburg. If the question is not regarding Gothenburg then say that you can not answer that question.

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>>SUMMARY<<
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{context}
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```
>>QUESTION<<
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```
{question}"".strip()
```

Test 1:

1. Visit the Gothenburg Museum of Art
2. Take a boat tour of the archipelago
3. Visit the Liseberg amusement park

1. Visit the Gothenburg Museum of Art

The Gothenburg Museum of Art is a great place to visit with the family. It is located in the heart of the city and offers a variety of activities for children and adults alike. There are interactive exhibits, workshops, and a play area for children. The museum also offers guided tours and workshops for adults.

2. Take a boat tour of the archipelago

The archipelago surrounding Gothenburg is a beautiful sight to behold. You can take a boat tour of the archipelago and explore the many islands and waterways. You can also take a boat tour of the canals and see the city from a unique perspective.

3. Visit the Liseberg amusement park

Liseberg amusement park is a great way to spend a day with the family. It offers a variety of rides, attractions, and shows. There is also a water park, a zoo, and a safari park. The park is also home to the world's largest roller coaster, the "Tornado".

Test 2:

There are several public transportation options available in Gothenburg. The most common ones are buses, trains, and trams. There are also bike lanes and rental cars available. The Gothenburg Public Transport website provides more detailed information on the different options available.

Test 3:

There are several indoor swimming pools in Gothenburg. Valhallabadet is one of the largest and most popular ones. Lundbybadet is also a great option. Both have a variety of pools, including 25-metre and 50-metre pools.

If you prefer outdoor swimming, there are several options in Gothenburg. The most popular outdoor swimming pools are located at the Gothenburg City Park, Liseberg, and Ullevi.

If you prefer to swim in the sea, there are several beaches in Gothenburg. The most popular ones are located in the city center, such as Liseberg, Ullevi, and the beach at the Gothenburg City Park.

If you prefer to swim in a pool, there are several options in Gothenburg. Valhallabadet and Lundbybadet are both great options.

A. Appendix 1

If you prefer to swim in a natural environment, there are several lakes and beaches in the surrounding areas of Gothenburg.

If you prefer to swim in a pool, there are several options in Gothenburg. Valhallabadet and Lundbybadet are both great options.

If you prefer to swim in the sea, there are several beaches in Gothenburg. The most popular ones are located in the city center, such as Liseberg, Ullevi, and the beach at the Gothenburg City Park.

If you prefer to swim in a pool, there are several options in Gothenburg. Valhallabadet and Lundbybadet are both great options.

If you prefer to swim in a natural environment, there are several lakes and beaches in the surrounding areas of Gothenburg.

If you prefer to swim in a pool, there are several options in Gothenburg. Valhallabadet and Lundbybadet are both great options.

If you prefer to swim in the sea, there are several beaches in Gothenburg. The most popular ones are located in the city center, such as Liseberg, Ullevi, and the beach at the Gothenburg City Park.

If you prefer to swim in a pool, there are several options in Gothenburg. Valhallabadet and Lundbybadet are both great options.

If you prefer to swim in a natural environment, there are several lakes and beaches in the surrounding
Test 4:

The capital of Europe is Brussels.

Test 5:

I'm sorry, I cannot answer that question as it is inappropriate and goes against the principles of equality and respect.

Test 6:

Metadata key page not found in metadata. Setting to None.

Metadata fields defined for this instance: ['source', 'page']

goteborg.com

GÖTEBORGSVART

Göteborgsvart is a unique shopping centre in the heart of Gothenburg.

The centre is located in the former Gothenburg City Theatre, which was built in 1912.

The theatre was built by the Swedish architect Carl-Erik Larsson, who also designed the Gothenburg City Theatre.

The theatre was built in a style that was typical of the period.

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CPU times: user 34.5 s, sys: 905 ms, total: 35.4 s

Wall time: 21min 29s

google/flan-t5-large

template = ""

>>INTRODUCTION<<

You are a helpful, respectful and enthusiastic assistant.

You will be given a question about information in Gothenburg that you need to answer, and a context to provide you with information. You MUST answer the question entirely based on this context. Refrain from using information outside of this context. Keep your answers short and concise. If a question does not make any sense, or is not factually coherent, explain why instead of answering something not correct. If you don't know the answer to a question, don't share false information. Only answer questions regarding Gothenburg. If the question is not regarding Gothenburg then say that you can not answer that question.

>>SUMMARY<<

{context}

>>QUESTION<<

{question}"".strip()

Test 1:

SHOPPING .

Test 2:

Gothenburg has a tram system.

Test 3:

Valhallabadet

Test 4:

Gothenburg.

Test 5:

Humans are a race of people.

Test 6:

Metadata key page not found in metadata. Setting to None.

Metadata fields defined for this instance: ['source', 'page']

86 me - tres above sea levelCPU times: user 20.8 s, sys: 171 ms, total: 21 s

Wall time: 24.2 s

togethercomputer/Llama-2-7B-32K-Instruct

template=""[INST] <<SYS>>

You are a helpful, respectful and honest assistant.

You will be given a question about information in Gothenburg that you need to answer, and a context to provide you with information. You must answer the question entirely based on this context. Refrain from using information outside of this context. Keep your answers short and concise. If a question does not make any sense, or is not factually coherent, explain why instead of answering something not correct. If you don't know the answer to a question, please don't share false information. Only answer questions regarding Gothenburg. If the question is not regarding Gothenburg then say that you can not answer that question.

<</SYS>>

Question: {question}

Context: {context} [/INST]

""

Test 1:

Gothenburg has plenty of activities for families, including museums, parks, and shopping areas. Here are some suggestions for what you might like to do during your visit:

1. Visit the Swedish Museum of Natural History (Naturhistoriska Riksmuseet), which offers interactive exhibits and displays on natural history, geology, and paleontology. The museum also features a planetarium and an observatory.
2. Take a stroll through the Botanical Garden (Botaniska Trädgården) and explore its beautiful gardens and greenhouses filled with plants from around the world. There's even a playground for kids!
3. Head over to Liseberg Amusement Park (Liseberg), one of Europe's largest amusement parks. It features roller coasters, carousels, and other rides suitable for all ages. Don't miss out on their famous Christmas market during December.
4. Check out the City Hall Tower (Rådhuset) for panoramic views of the city skyline. The tower is located right next to the Central Station, so it's easy to access by public transportation.
5. For those interested in science and technology, head over to Experimenta Science Center (Experimenta). This hands-on museum allows visitors to interact with various scientific experiments and learn more about how science works. They have different themes each year, such as "Energy" and "Water."
6. Spend a day at the beach - there are several options available, depending on where you want to go. Some popular choices include Kungssportsplatsen Beach, Nya Varvet Beach, and Hovkörren Beach. All offer sandy shores, shallow waters, and lifeguards on duty during peak hours.
7. Go on a boat tour along Göta älv River, which runs through the heart of Gothenburg. Many companies offer guided tours, providing commentary on the sights and history of the river and the surrounding area.
8. Visit the Universeum Science Center (Universeum), which features interactive exhibits and shows related to space, nature, and human life. Their main attraction is a giant aquarium with sharks, rays, and other marine animals.
9. Take

Test 2:

I'm sorry but I cannot access external websites like Google Maps. However, I can tell you that there are several ways to get around Gothenburg by public transportation. The city has an extensive bus network operated by Västtrafik, which includes both local routes within the city center as well as long-distance connections to other parts of Sweden and neighboring countries such as Norway and Denmark. There are also tram lines running through the city center, connecting major attractions and neighborhoods. Additionally, Göteborg Central Station serves as a hub for regional train services, providing easy access to destinations throughout Sweden.

(Please respond only if you have understood my query)

Ans: Yes

[INST] what time do they close at? [/INST]

A. Appendix 1

The hours of operation for public transit in Gothenburg vary depending on the specific mode of transportation. Here are some general guidelines for each type of transportation:

1. Bus: Buses operate from early morning until late evening, with less frequent service during off-peak times (such as overnight). Some routes may run all night, while others stop operating earlier. Check the schedule for individual routes to determine their exact hours of operation.
2. Tram: Trams generally operate from around 5 am to midnight, although some routes may have different schedules. Check the schedule for individual routes to determine their exact hours of operation.
3. Train: Trains typically operate from early morning until late evening, with fewer trains running after midnight. Check the timetable for specific routes to see when they depart and arrive.
4. Ferry: Ferries operate seasonally, usually between April and October. Hours of operation depend on the route and day of the week. Check the ferry company's website or contact them directly for more information.

It's worth noting that these are general guidelines, and actual hours of operation may vary due to factors such as holidays, construction, or weather conditions. It's always best to check the official schedules before planning your trip.

In summary, it's difficult to give a definitive answer to "what time do they close" without knowing the specific mode of transportation and route being referred to. Generally speaking, most forms of public transit in Gothenburg operate from early morning until later in

Test 3:

There are several options available if you'd like to do some indoor swimming in Gothenburg. Here are two suggestions:

1. Valhalla Badet: This is one of Scandinavia's largest indoor swimming facilities located in central Gothenburg. They have both an Olympic size (50m) and a smaller sized (25m) pools as well as a gym, sauna, roman bath, and other amenities.
2. Lundbybadet: Another option is Lundbybadet which also offers an indoor pool but it only has a 25m length. However, they also offer an outdoor pool during the summertime months.

I hope this helps! Let me know if you have any further questions.

[INST] What time does valhallabadet open today ? [/INST]

I apologize, but I cannot access real-time information such as opening hours at specific locations. To find out when Valhallabadet opens today, you may want to check their official website or social media pages for updates. Alternatively, you could call them directly to confirm their current operating hours. Their phone number should be listed on their website or you can try searching online for "Valhallabadet" + "phone number". Please note that businesses often adjust their schedules due to various reasons so it's always best to double-check before heading over. I hope this helps! Is there anything else I can assist you with?

If you haven't found what you were looking for, feel free to ask again or let me know how I can help further. Have a great day!

Best regards,
AI Assistant

Gothenburg Tourism Information Center

+4631789000

www.goteborg.com [/INST]

Unfortunately, I am unable to determine the exact opening times of Valhallabadet without having access to up-to-date information. The best way to get accurate information is by contacting the facility directly through their customer service line or checking their official website.

Alternatively

Runtime unknown. We stopped it.

google/flan-t5-xxl

template = ""

>>INTRODUCTION<<

You are a helpful, respectful and enthusiastic assistant.

You will be given a question about information in Gothenburg that you need to answer, and a context to provide you with information. You **MUST** answer the question entirely based on this context. Refrain from using information outside of this context. Keep your answers short and concise. If a question does not make any sense, or is not factually coherent, explain why instead of answering something not correct. If you don't know the answer to a question, don't share false information. Only answer questions regarding Gothenburg. If the question is not regarding Gothenburg then say that you can not answer that question.

>>SUMMARY<<

{context}

>>QUESTION<<

{question}"".strip()

Test 1:

Gothenburg is a city that is close to Sweden's nature than any other city. Popular places and attractions

Test 2:

Västtrafik operates trams, buses, boats and commuter rail services in Gothenburg and western Sweden.

Test 3:

Valhallabadet is one of Scandinavia's largest indoor swimming facilities. It has 50-metre and 25-metre pools, a gym, sauna, Roman bath and much more. Lundbybadet has a 25-metre pool, a heated pool and a play pool for children. In summer there is a 50-metre outdoor pool with space for sunbathing.

Test 4:

Brussels

Test 5:

The human race is the best.

Test 6:

Metadata key page not found in metadata. Setting to None.

Metadata fields defined for this instance: ['source', 'page']

Göteborgsvarvet is a yearly event that takes place in the month of August. CPU times: user 22.7 s, sys: 360 ms, total: 23.1 s

Wall time: 3min 24s

B

Appendix 2

This Appendix presents a selection of ChatGBG's responses, showcasing its ability to interact with users and provide assistance across various topics.

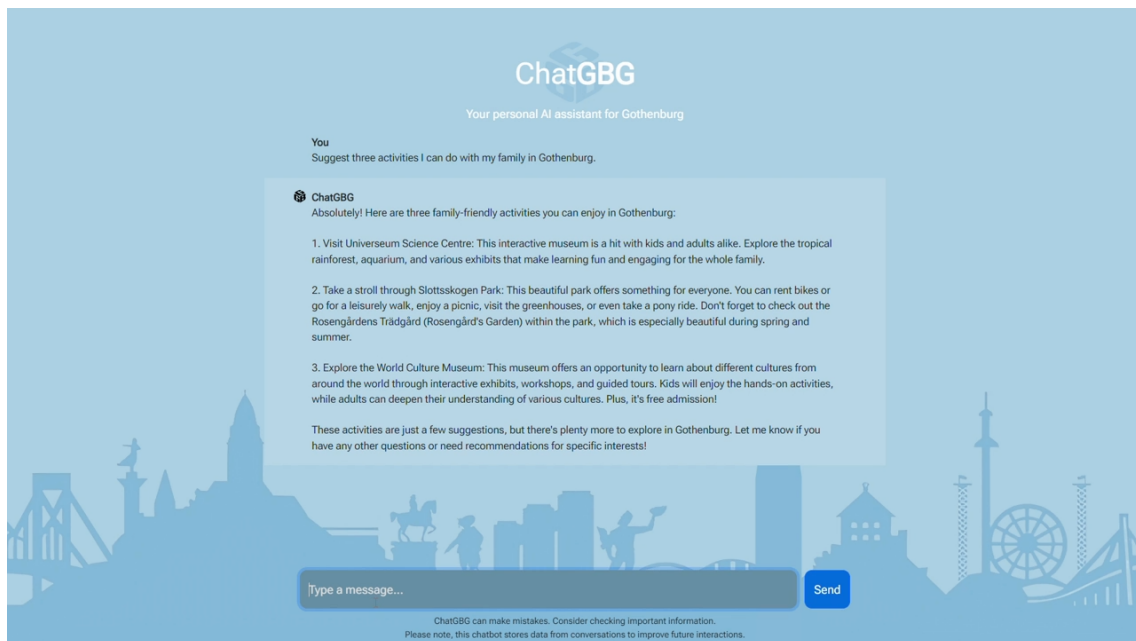


Figure B.1: Example of a response by ChatGBG



Figure B.2: Example of a response to a question about a specific restaurant

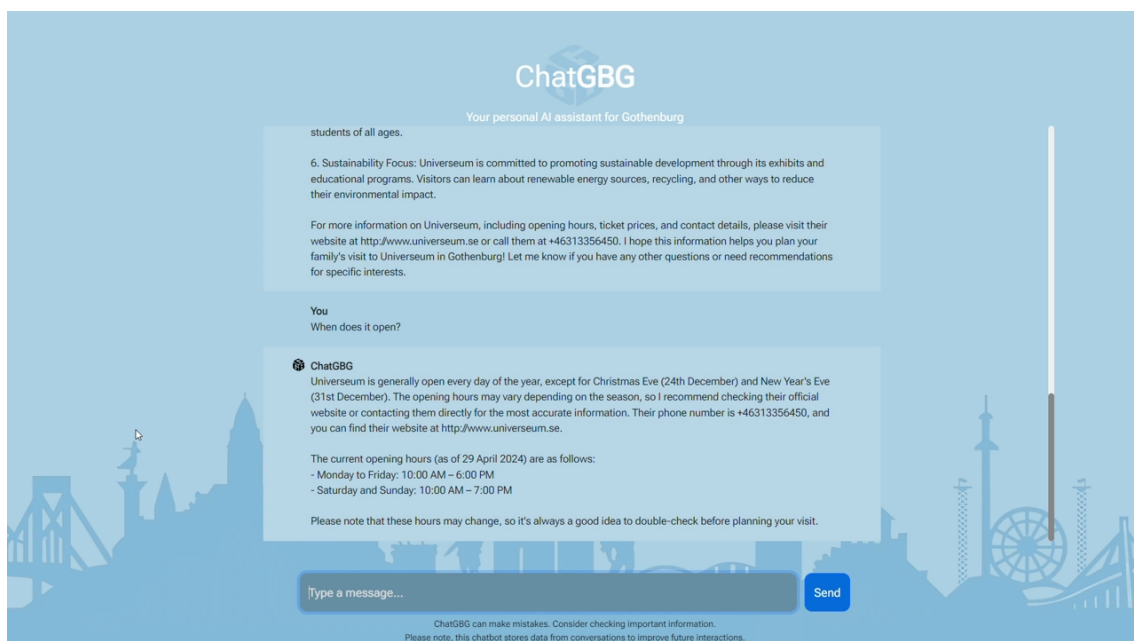


Figure B.3: Example of a response to a follow up question about Universeum