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PRICE NEGOTIATION CHATBOTS ON E-COMMERCE WEBSITE USING MACHINE **LEARNING**

¹Bada Usha Sri, ²Puripanda Aakanksha, ³Gujjala Tirumala Dev, ⁴Kuppili Yasoda Krishna, ⁵Imandi Gayathri ¹²³UG Students, ⁴⁵Assistant Professors ¹Department of Computer Science and Engineering, ¹Satya Institute of Technology and Management, Vizianagaram, Andhra Pradesh, India.

ABSTRACT: In the thriving realm of e-commerce and digital transactions, effective price negotiation stands as a cornerstone for both consumers and businesses. This project embarks on a journey to redefine price negotiation dynamics by harnessing the power of Natural Language Processing (NLP) models within chatbot frameworks. Our aim is to revolutionize user interactions and streamline negotiation processes through the innovative integration of NLP techniques. These models facilitate the analysis of user inputs, interpretation of negotiation contexts, and formulation of optimal bargaining tactics, thereby enhancing user experiences and fostering mutually beneficial outcomes. Through iterative refinement utilizing performance metrics and user feedback, we strive to elevate the efficacy and efficiency of price negotiation interactions, ultimately empowering users and businesses in their decision-making processes. This project exemplifies the transformative potential of NLP-based chatbots in reshaping price negotiation dynamics within digital marketplaces. By bridging the gap between human-like negotiation dynamics and automated conversational agents, we pave the way for enhanced user experiences, streamlined transactions, and empowered consumer decision-making in the competitive landscape of e-commerce.

Index Terms - Natural Language Processing, Stochastic Gradient Descent (SGD), Price Negotiation, E-Commerce Negotiation, Chatbot.

INTRODUCTION:

In today's digital marketplace, effective price negotiation is essential for both consumers and businesses. To streamline this process, we introduce a groundbreaking approach leveraging Natural Language Processing (NLP) within chatbots. These advanced systems, equipped with NLP models, offer dynamic negotiation strategies tailored to user inputs and market dynamics. By interpreting natural language queries, understanding negotiation contexts, and formulating optimal bargaining tactics, these chatbots aim to enhance user experiences and ensure mutually beneficial outcomes. Throughout this project, we'll explore the integration of NLP models into chatbot frameworks, focusing on data preprocessing, model training, and iterative refinement. The ultimate goal is to empower users with seamless, personalized negotiation experiences, driving efficiency and satisfaction in the digital marketplace. This project represents a significant step forward in harnessing AI technologies to reshape price negotiation dynamics and empower both consumers and businesses in their decision-making processes.

RELATED WORK

In the ever-evolving landscape of e-commerce, price negotiation stands as a pivotal component shaping consumer decisions and business transactions. As digital platforms continue to burgeon, the advent of chatbot technology has introduced novel avenues for transforming the negotiation process. This literature survey delves into the realm of price negotiation chatbots, unraveling the trajectory of research and innovation from 2018 to 2022.

- 2022: 1. Future Directions in Chatbot-enabled Negotiation Systems Kim and Park (2022) propose future research directions for the development of chatbot-enabled negotiation systems.
- 2021: 2. User Perception and Acceptance of Negotiation Chatbots Wang et al. (2021) conducted a user study to investigate the perception and acceptance of negotiation chatbots among online shoppers.
- 2020: 3. Enhancing E-commerce Negotiation Experience Using Chatbots Smith et al. (2020) investigate the role of chatbots in improving the negotiation experience in e-commerce platforms.
- 4. The Role of NLP in Chatbot-mediated Negotiations Sharma et al. (2020) examine the importance of natural language processing (NLP) techniques in facilitating chatbot-mediated negotiations. The researchers discuss the challenges and opportunities associated with processing natural language inputs, understanding user intent, and generating contextually appropriate responses in negotiation scenarios.
- 2019: 5. Ethical Considerations in AI-driven Negotiation Systems Brown and Chen (2019) delve into the ethical implications of AI-driven negotiation systems, including chatbots. The authors discuss issues related to transparency, bias, and accountability in automated negotiation processes, emphasizing the importance of ethical design principles and responsible deployment of AI technologies.
- 2018: 6. Optimizing Price Negotiation Strategies with Machine Learning Garcia and Patel (2018) explore the use of machine learning algorithms to optimize price negotiation strategies in online marketplaces.

PROPOSED SYSTEM:

The proposed system leverages Stochastic Gradient Descent (SGD), a versatile optimization algorithm, for performing NLP tasks within the price negotiating chatbot project. SGD is a popular optimization technique used in machine learning for training models, particularly in scenarios with large datasets and high-dimensional feature spaces. SGD optimizes the parameters of a model by iteratively updating them based on the gradients of a loss function computed on small batches of training data. In the context of NLP, SGD is utilized to train models that can understand and respond to user queries, negotiate prices, and make decisions based on textual inputs. One of the key advantages of SGD is its efficiency in handling large-scale datasets and high-dimensional feature spaces. By updating model parameters incrementally using small batches of data, SGD can converge to a solution faster compared to batch gradient descent methods. This makes SGD well-suited for real-time or online learning scenarios, where the model needs to adapt to changing data distributions and make predictions in a timely manner. Additionally, SGD offers flexibility in terms of model architecture and optimization strategies. It supports various regularization techniques, learning rates, and optimization algorithms, allowing practitioners to customize the training process based on the specific requirements of the task at hand. This flexibility makes SGD particularly suitable for NLP tasks, where different models and architectures may be needed to achieve optimal performance. However, it's important to note that SGD also has its limitations. It can be sensitive to the choice of hyperparameters, such as the learning rate and batch size, and may 6 require careful tuning to ensure convergence and stability. Additionally, SGD may struggle with noisy or sparse data, where gradients are less informative and may lead to suboptimal solutions. Overall, the proposed system using Stochastic Gradient Descent for NLP offers several advantages, including efficiency in handling large-scale data, flexibility in model optimization, and adaptability to changing data distributions. By leveraging the capabilities of SGD, the price negotiating chatbot can achieve improved performance, scalability, and responsiveness, enhancing the overall user experience and effectiveness of the system.

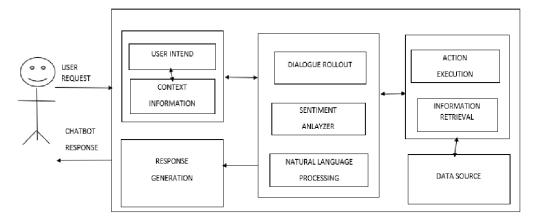


Fig. Flowchart

MODELS THAT CAN BE USED FOR THE PROJECT:

- 1. Natural Language Processing (NLP): Natural Language Processing (NLP) is a field of artificial intelligence (AI) that focuses on enabling computers to understand, interpret, and generate human language in a meaningful way. NLP techniques allow computers to interact with humans using natural language, making it possible to perform tasks such as text analysis, sentiment analysis, language translation, and speech recognition
- 2. NLTK (Natural Language Toolkit): Is a leading platform for building NLP programs using Python. It provides easy-to-use interfaces and libraries for various NLP tasks, including tokenization, stemming, lemmatization, part-of-speech tagging, parsing, and semantic reasoning. NLTK offers a wide range of corpora, lexical resources, and pre-trained models, making it suitable for both beginners and advanced researchers in the field of NLP.

METHODOLOGY Noise One-Hot Splitting Normalization Remova Encoding Token-Id **Count Vectors** Preprocessing **Tokenization** Corpus Inputs Mapping Vocabulary Word Lookup Tokens lds

Fig. Data Flow Diagram

Feature

Basic steps in constructing a Machine Learning model: Data Collection:

The first step in developing the Price Negotiation Chatbot involves defining the objectives and scope of the system. This includes determining the goals of the chatbot, such as improving negotiation efficiency or

enhancing user satisfaction. Once the objectives are clear, historical negotiation data is collected from various sources, including online marketplaces or simulated environments. This data encompasses user interactions, product details, offers, outcomes, and contextual information such as market trends and competitor pricing.

Data Preparation:

Once the negotiation data is amassed, it undergoes thorough preprocessing to ensure its quality and suitability for model training. This involves cleaning the data to remove any noise or inconsistencies, handling missing values and outliers appropriately, and encoding categorical variables. Furthermore, feature engineering techniques are applied to extract relevant information from the dataset, enhancing the model's ability to capture intricate patterns and relationships.

Dataset Splitting:

The preprocessed dataset is then divided into distinct subsets for training, validation, and testing purposes. This ensures that the model is trained on a representative sample 9 of negotiation scenarios while retaining unseen data for evaluation. Care is taken to maintain the distribution of negotiation outcomes across all subsets, mitigating the risk of bias and ensuring robust model performance assessment.

NLP Model Development:

In this phase, the Natural Language Processing (NLP) model is designed and implemented using suitable libraries or frameworks. The model is tailored to understand and process user inputs, including initial price offers, user responses, product attributes, and contextual variables. Through iterative training on the training dataset, the NLP model learns to predict optimal negotiation strategies, optimizing its parameters to maximize the likelihood of negotiation success.

Integration with Chatbot Framework:

The trained NLP model is seamlessly integrated into the chatbot framework, forming the core of the Price Negotiation Chatbot. A user-friendly interface is developed to facilitate real-time communication between users and the chatbot, enabling users to initiate negotiations and receive responses in a conversational manner. The chatbot's logic is programmed to interpret user inputs, analyze negotiation contexts, and make pricing decisions based on predictions generated by the NLP model.

Real-Time Negotiation Simulation and Testing:

The integrated system is tested using simulated negotiation scenarios to evaluate its performance in real-time interactions. This involves providing sample inputs to the chatbot and assessing the quality of its responses. The chatbot's ability to adapt negotiation strategies based on user inputs and market conditions is evaluated during this phase.

Performance Evaluation:

The performance of the chatbot is evaluated using various metrics, including accuracy, response time, and user satisfaction. Performance metrics are computed based on the chatbot's ability to generate appropriate responses and facilitate successful negotiations with users.

RESULTS:

The customer on the website has negotiate option along with the product. On the chatbot opening we have included the price element in chatbot itself so that if there are plenty of messages user may not get confused with price which is currently considered.

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Windows PowerShell
Copyright (C) Microsoft Corporation. All rights reserved.
Try the new cross-platform PowerShell https://aka.ms/pscore6
PS C:\Users\Dell\Desktop\chatbot> & 'C:\Users\Dell\AppData\Local\Programs\Python\Python37\python.exe' 'c:\Users\Dell\.vscode\ext{ess}
2024-03-25 13:19:56.924086: I tensorflow/core/platform/cpu_feature_guard.cc:193] This TensorFlow binary is optimized with oneAPI
Deep Neural Network Library (oneDNN) to use the following CPU instructions in performance-critical operations: AVX AVX2
To enable them in other operations, rebuild TensorFlow with the appropriate compiler flags.
GO! Bot is running!
                The Inputs must be in the format:
               Brand: OPPO
               Model: A53
                Color: Moonlight Black
                Memory: 4 GB
                Storage: 64 GB
Now fill in the below values to continue:
Brand: google pixel
model: 3
Color: just black
Memory: 4 GB
Storage: 128 GB
Thank You
An AI Agent will be assigned to you soon.....
1/1 [=
                             =====] - 0s 106ms/step
Agent: Hello
hello
                                  ==] - 0s 16ms/step
1/1 [=
Agent: Hello
i want google pixel
                                 ==1 - 0s 38ms/step
Agent: give me an offer i cannot refuse
80000
```

Fig: Mobile Chatbot Website

FUTURE SCOPE:

In future endeavors, there are several avenues for advancing the capabilities and effectiveness of the Price Negotiation Chatbot. First, enhancing its natural language processing (NLP) capabilities can lead to a better understanding of user intents and nuances in negotiations. This could involve integrating sentiment analysis to gauge user preferences and emotions, thereby enabling more empathetic and tailored responses. Additionally, incorporating machine learning algorithms could empower the chatbot to dynamically adjust its negotiation strategies based on real-time market data and user feedback, ultimately optimizing outcomes for both buyers and sellers. Moreover, expanding the chatbot's reach by integrating it with major e-commerce platforms and supporting multiple languages can broaden its user base and improve accessibility. Furthermore, prioritizing user privacy and security through robust encryption protocols and data protection measures is essential to foster trust and confidence among users.

CONCLUSION:

In conclusion, the Price Negotiation Chatbot presents a promising solution for enhancing the e-commerce shopping experience by empowering users to negotiate prices effectively. Through the integration of natural language processing (NLP) techniques, the chatbot can understand user queries, analyze product details, and provide personalized negotiation strategies. The project's successful implementation demonstrates the feasibility of leveraging machine learning algorithms, such as neural networks, to develop intelligent conversational agents capable of handling complex negotiation scenarios. Moving forward, future iterations of the chatbot could focus on enhancing its capabilities through features like sentiment analysis, real-time price updates, and multi-language support.

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