

Milestone 1 Report

BankBot AI Chatbot for Banking FAQs

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Batch: 5

Objective:

- **Intent Classification:** Classifying user queries into predefined intents.
- **Entity Extraction:** Extracting key entities from the queries (e.g, account numbers, dates).
- **Response Generation:** Providing a predefined response based on the predicted intent.

Tasks Completed:

1. Dataset Loading and Preprocessing:

- Loaded a CSV file containing columns: text (query), intent (predicted intent), response (response for the intent), and entities (optional).
- Checked for missing required columns (text, intent, response) and ensured the dataset was complete.

2. Entity Parsing:

- Implemented a function (safe_parse_entities) to handle the entities column, ensuring safe parsing of different formats. This function handles missing or malformed entities gracefully.

3. Intent Classification:

- Prepared the data for training by splitting it into X (input text) and y (intent label).
- Used **TfidfVectorizer** for converting text into numerical features (tf-idf scores).
- Applied a **Logistic Regression** model to classify the intent based on the vectorized text.
- The classifier was trained using 80% of the dataset (train-test split) and tested on the remaining 20% to evaluate its performance.

4. Model Evaluation:

- Evaluated the classifier's performance using the **classification report**, which provides metrics such as precision, recall, and F1-score for each intent class.

== Classification Report ==					
	precision	recall	f1-score	support	
block_card	0.00	0.00	0.00	1	
branch_locator	0.00	0.00	0.00	1	
card_inquiry	0.00	0.00	0.00	1	
check_balance	1.00	1.00	1.00	2	
goodbye	1.00	1.00	1.00	1	
loan_inquiry	1.00	1.00	1.00	2	
out_of_scope	0.00	0.00	0.00	1	
transfer_money	0.20	1.00	0.33	1	
accuracy			0.60	10	
macro avg	0.40	0.50	0.42	10	
weighted avg	0.52	0.60	0.53	10	

5. Entity Extraction:

- Integrated spaCy for Named Entity Recognition (NER) to extract entities from the text.
- A custom rule was added to extract numeric tokens as account_number, assuming account numbers are typically numeric.
- Extracted entities are stored in a dictionary where the key is the entity type (e.g., account_number, ORG, GPE), and the value is the entity's text.

6. Response Retrieval:

- A function (get_response) was implemented to retrieve the corresponding response from the dataset based on the predicted intent.
- The system returns a predefined response when a valid intent is predicted; if no response is found for the predicted intent, a default message is returned.

Tools and Libraries Used

- **Pandas**: For handling the dataset and performing data cleaning.
- **spaCy**: For Named Entity Recognition (NER) and custom entity extraction.
- **Scikit-learn**: For the machine learning pipeline, including:
 - **TfidfVectorizer** for text vectorization.
 - **Logistic Regression** for intent classification.
 - **train_test_split** for splitting data into training and testing sets.
 - **classification_report** for model evaluation.

Conclusion:

Milestone 1 provided a solid foundation for building a conversational AI system that classifies user intents, extracts relevant entities, and generates appropriate responses. Key skills learned include data preprocessing, feature extraction using tf-idf, intent classification with Logistic Regression, and named entity extraction with spaCy. The system is functional but can be improved by tuning the model and expanding entity recognition capabilities.