



Phase-2SubmissionTemplate

Student Name: [JAGADEESHWARI J]

RegisterNumber:[422723104043]

Institution: [V.R.S. College of engineering and

technology]

De partment:[Computer Science and

Engineering]

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Topic: [Revolutionizing customer support with an intelligent chatbot for automated assistance]

GithubRe positoryLink:htt ps:llgithub.comlJagad eeshwari279lJagadeeshwari.git

1. Problem Statement

This project aims to develop an Alpowered chatbot capable of understanding customer queries
and responding with relevant, accurate information. The
problem type is primarily a combination of:

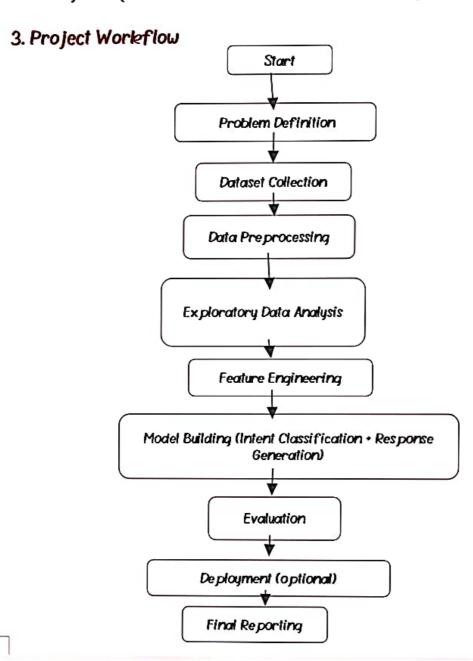




- Text classification (intent detection)
- Mamed Entity Recognition (NER)
- Sequence generation or retrieval (response generation)

2. Project Objective

- Technical Goal: Build a machine learningINLP-based chatbot that can classify intents, extract entities, and provide automated, accurate replies.
- Model Objective: High classification accuracy for intent detection and response relevance for out put generation.
- Post-EDA Refinement: Identified the need for better handling of ambiguous queries and added fallback handling.







4. Data Description

- FLink: https://www.kaggle.com/datasets/niraliivaghani/chatbot-dataset
- Type of Data : unstructured
- Mumber of Records and Features: -100,000 records with fields like user query,
 intent label, timestamp, and response
- **TACK**
- fraget Variable: Intent category (e.g., billing issue, login failure, general inquiry)

5. Data Preprocessing

Link:https://colab.research.google.com/drivel/fmbgnwcCdFTD80C4UagdaWLNaGg
J2_E3?usp=sharing

5.1 Handling values:

Handling Missing Values : Removed incomplete conversations

5.2 Removing or Justifying Duplicate Records

Duplicates: Dropped repeated entries

5.3 Text Cleaning

Lowercasing, punctuation removal, stopword removal, lemmatization labels

Encoding: Label encoding for intent labels

5.4 Encoding:

Label encoding for intent labels

5.5 Vectorization

TF-IDF I word embeddings (e.g., GloVe or BERT)





5.6 Normalization

Not required for text, but token length truncation applied.

6. Exploratory Data Analysis (EDA)

- GUnivariate: Word clouds and frequency plots for top words in each intent
- <u>Multivariate</u>: Count plots by intent class analysis of average query length by intent

7. Feature Engineering

- Custom Features: Message length, number of keywords matched
- MLP features: TF-IDF scores, POS tags
- **Tustification**: Enhanced model's ability to separate intents using semantic clues

8. Model Building

Models Used:

- Logistic Regression ((baseline classifier))
- FBERT-based fine-tuned Transformer for intent classification

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- **←Split**: 80/20 train-test split using train_test_split.
- Metrics: Accuracy, Precision, Recall, F1-score
- 9. Visualization of Results & Model Insights
- Confusion matrix for intent classification





- FROC Curve (if applicable)
- ft-SNE plot for visualizing intent separability
- Top influential words/features per intent class

 Findings
- FBERT achieved F1-score > 90%

Errors occurred mainly in overlapping or ambiguous intents

10. Tools and Technologies Used

- **Canguage:** Python
- FLibraries: pandas, numpy, scikit-learn, NLTK, spaCy, Transformers
- **Transfer of the Proof of the P**

11. Team Members and Roles

JAGADEESHWARI J - Documentation and visualization

JAMUNA RANI V- Feature engineering

ISHWARIYA A - Model development

JAYABHARATHI J - Data cleaning, EDA